

Network Systems  
Science & Advanced  
Computing  
Biocomplexity Institute  
& Initiative  
University of Virginia

# Estimation of COVID-19 Impact in Virginia

September 15<sup>th</sup>, 2021

(data current to September 11<sup>th</sup> -14<sup>th</sup>)

Biocomplexity Institute Technical report: TR 2021-101



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**BIOCOMPLEXITY** INSTITUTE

[biocomplexity.virginia.edu](https://biocomplexity.virginia.edu)

# About Us

- Biocomplexity Institute at the University of Virginia
  - Using big data and simulations to understand massively interactive systems and solve societal problems
- Over 20 years of crafting and analyzing infectious disease models
  - Pandemic response for Influenza, Ebola, Zika, and others



## Points of Contact

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# Overview

- **Goal:** Understand impact of COVID-19 mitigations in Virginia
- **Approach:**
  - Calibrate explanatory mechanistic model to observed cases
  - Project based on scenarios for next 4 months
  - Consider a range of possible mitigation effects in "what-if" scenarios
- **Outcomes:**
  - Ill, Confirmed, Hospitalized, ICU, Ventilated, Death
  - Geographic spread over time, case counts, healthcare burdens

# Key Takeaways

Projecting future cases precisely is impossible and unnecessary.

Even without perfect projections, we can confidently draw conclusions:

- **Case rate growth in Virginia has slowed as many districts have started a decline from a peak, similarly across US many states have started declining from a peak; Case rates remain very high.**
- VA mean 7-day incidence is up at 43/100K from 38/100K, US also up at 50/100K (from 48/100K)
- Projections show reduced rate of increase and thus the impact has decreased compared to last week
- Recent updates:
  - Adjustment to higher levels of assumed immunity waning (natural and vaccine)
  - Added a SeptSurge based on transmission rates from last year Labor Day to Thanksgiving with variant boosting
  - Added Fall surge scenario to capture potential rebounds and further test immunity from expanded vaccination

The situation continues to change. Models continue to be updated regularly.



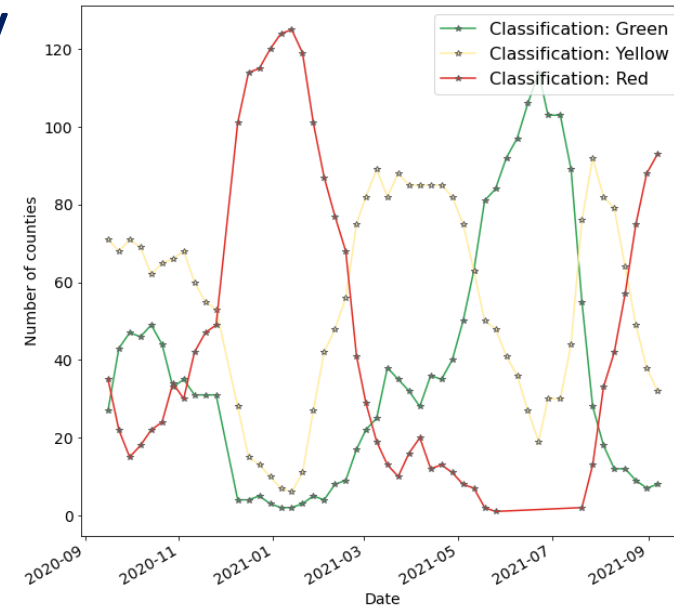
# Situation Assessment

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# Case Rates (per 100k) and Test Positivity

- Case rate increase across all health districts
- Some past 50% of winter peak and growing
- More than 50% of counties with TPR > 10%

Data source: <https://data.cms.gov/covid-19/covid-19-nursing-home-data>



## County level RT-PCR test positivity

**Green:** <5.0% (or <20 tests in past 14 days)

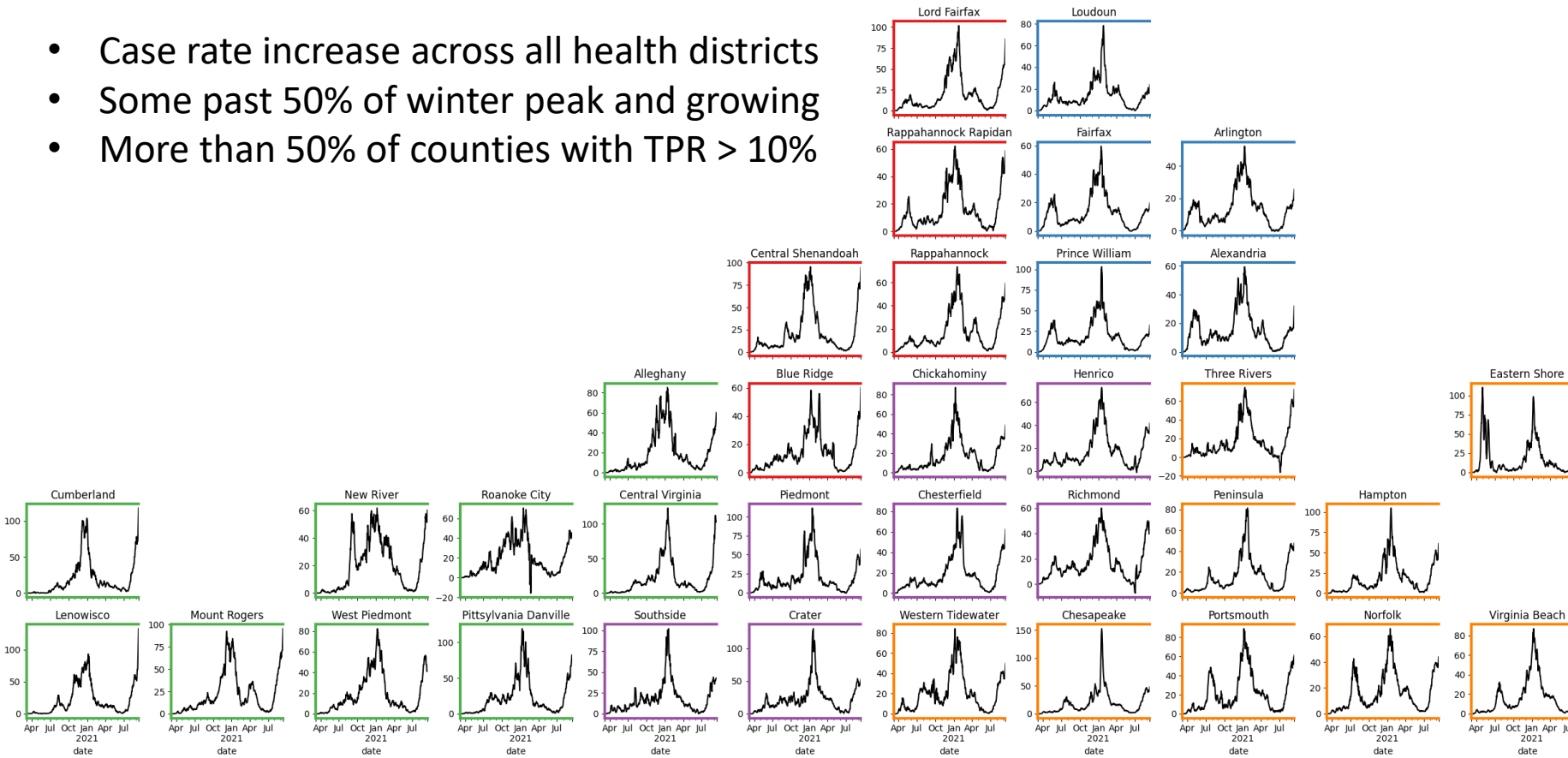
**Yellow:** 5.0%-10.0% (or <500 tests and <2000 tests/100k and >10% positivity over 14 days)

**Red:** >10.0% (and not "Green" or "Yellow")

Classification Green Yellow Red

date

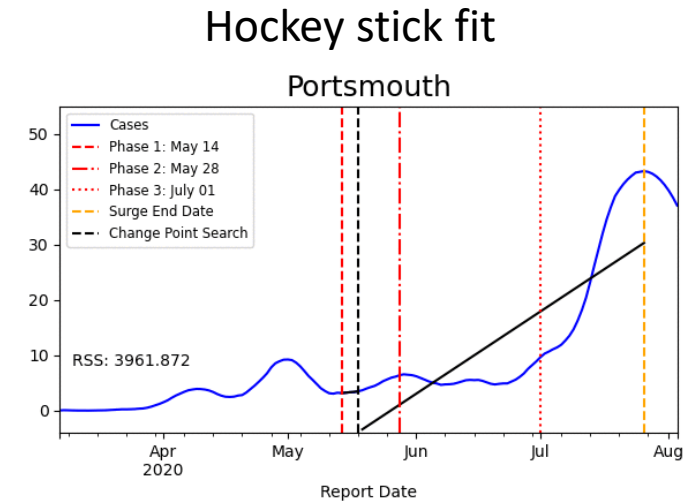
|            |       |      |      |
|------------|-------|------|------|
| 2021-06-28 | 103.0 | 30.0 | 0.0  |
| 2021-07-06 | 103.0 | 30.0 | 0.0  |
| 2021-07-13 | 89.0  | 44.0 | 0.0  |
| 2021-07-20 | 55.0  | 76.0 | 2.0  |
| 2021-07-27 | 28.0  | 92.0 | 13.0 |
| 2021-08-03 | 18.0  | 82.0 | 33.0 |
| 2021-08-10 | 12.0  | 79.0 | 42.0 |
| 2021-08-17 | 12.0  | 64.0 | 57.0 |
| 2021-08-24 | 9.0   | 49.0 | 75.0 |
| 2021-08-31 | 7.0   | 38.0 | 88.0 |



# District Trajectories

**Goal:** Define epochs of a Health District's COVID-19 incidence to characterize the current trajectory

**Method:** Find recent peak and use hockey stick fit to find inflection point afterwards, then use this period's slope to define the trajectory

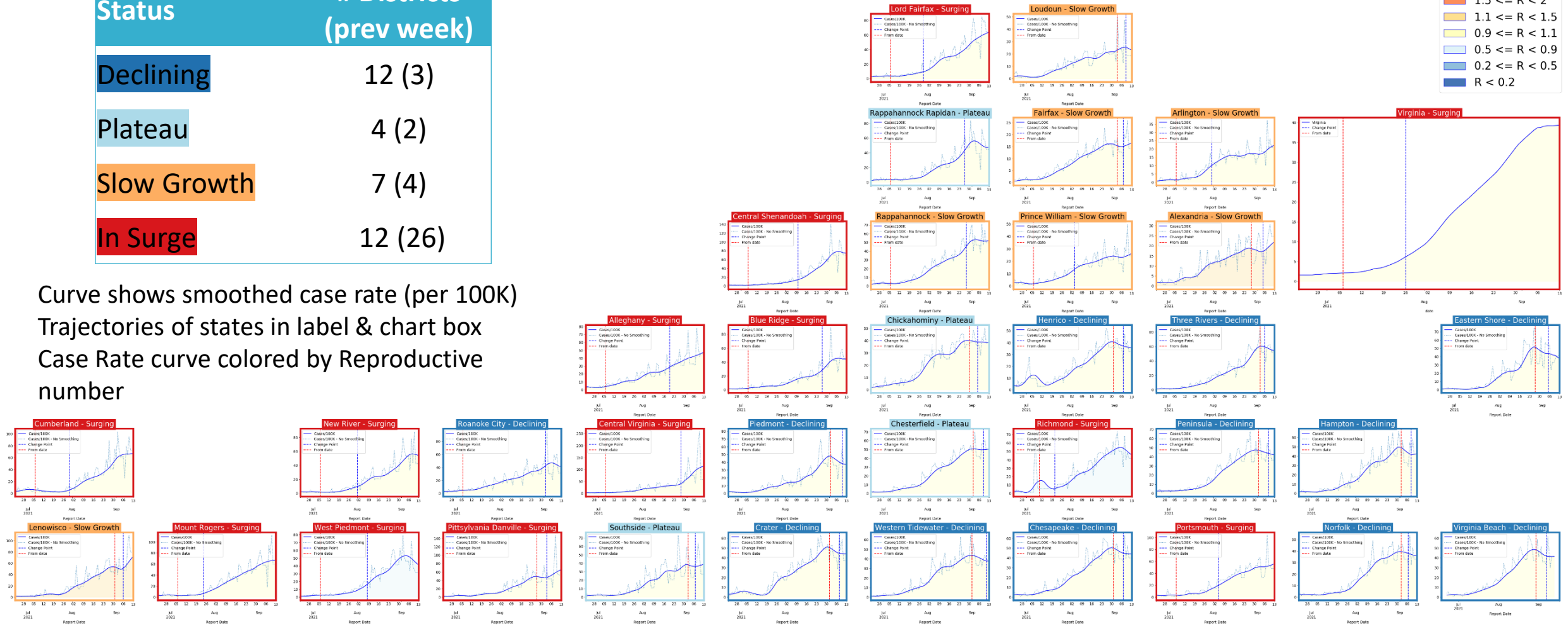
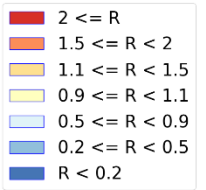


| Trajectory         | Description   | Weekly Case Rate (per 100K) bounds | # Districts (prev week) |
|--------------------|---|------------------------------------|-------------------------|
| <b>Declining</b>   | Sustained decreases following a recent peak                   | below -0.9                         | 12 (3)                  |
| <b>Plateau</b>     | Steady level with minimal trend up or down                    | above -0.9 and below 0.5           | 4 (2)                   |
| <b>Slow Growth</b> | Sustained growth not rapid enough to be considered a Surge    | above 0.5 and below 2.5            | 7 (4)                   |
| <b>In Surge</b>    | Currently experiencing sustained rapid and significant growth | 2.5 or greater                     | 12 (26)                 |

# District Trajectories – last 10 weeks

| Status      | # Districts<br>(prev week) |
|-------------|----------------------------|
| Declining   | 12 (3)                     |
| Plateau     | 4 (2)                      |
| Slow Growth | 7 (4)                      |
| In Surge    | 12 (26)                    |

Curve shows smoothed case rate (per 100K)  
Trajectories of states in label & chart box  
Case Rate curve colored by Reproductive  
number



# Estimating Daily Reproductive Number

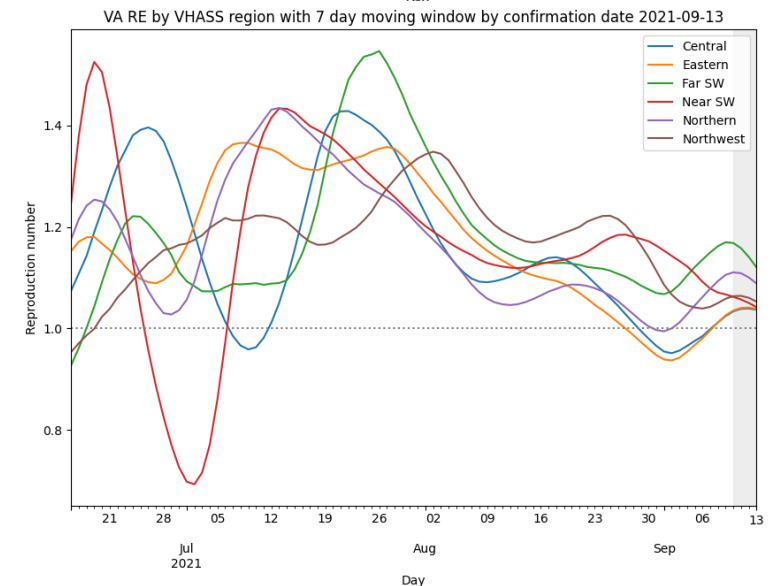
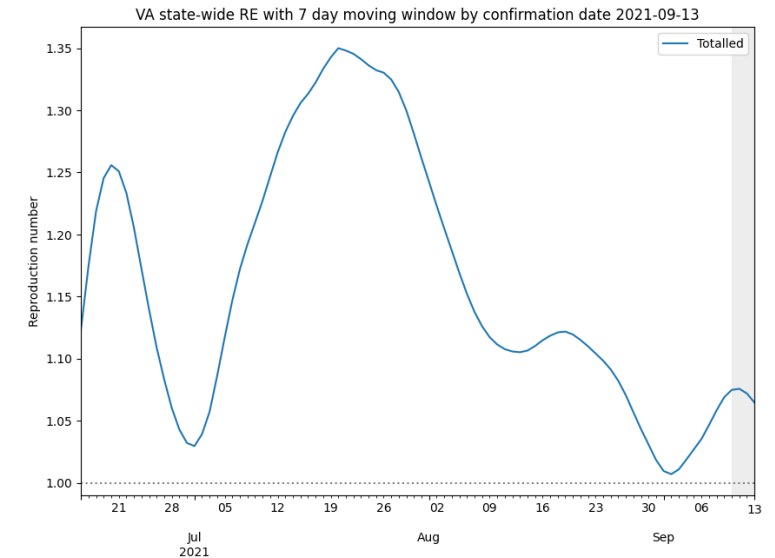
## Sept 13<sup>th</sup> Estimates

| Region     | Date Confirmed $R_e$ | Date Confirmed Diff Last Week |
|------------|----------------------|-------------------------------|
| State-wide | 1.065                | 0.062                         |
| Central    | 1.037                | 0.129                         |
| Eastern    | 1.039                | 0.134                         |
| Far SW     | 1.121                | 0.184                         |
| Near SW    | 1.042                | 0.023                         |
| Northern   | 1.089                | 0.174                         |
| Northwest  | 1.053                | 0.071                         |

### Methodology

- Wallinga-Teunis method (EpiEstim<sup>1</sup>) for cases by confirmation date
- Serial interval: updated to discrete distribution from observations (mean=4.3, Flaxman et al, Nature 2020)
- Using Confirmation date since due to increasingly unstable estimates from onset date due to backfill

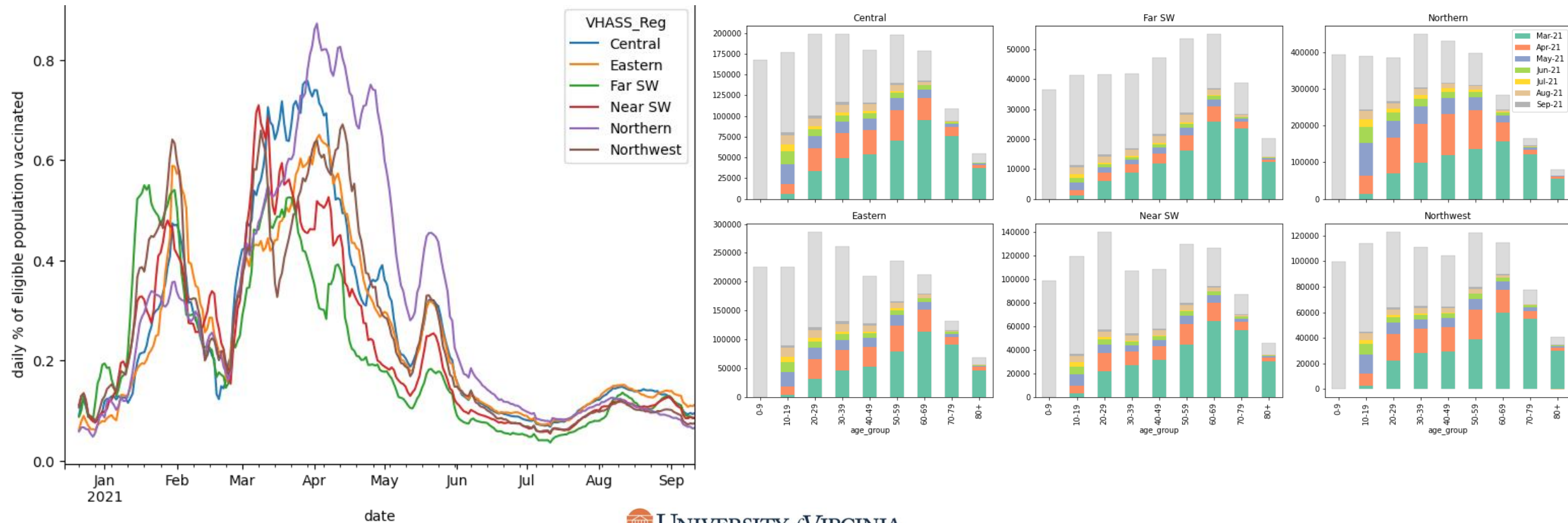
1. Anne Cori, Neil M. Ferguson, Christophe Fraser, Simon Cauchemez. A New Framework and Software to Estimate Time-Varying Reproduction Numbers During Epidemics. American Journal of Epidemiology, Volume 178, Issue 9, 1 November 2013, Pages 1505–1512, <https://doi.org/10.1093/aje/kwt133>



# Vaccination Administration Slows

## Regional Vaccine courses initiated per day:

- Total counts of first dose of vaccines across regions
- Age-specific proportions of population vaccinated

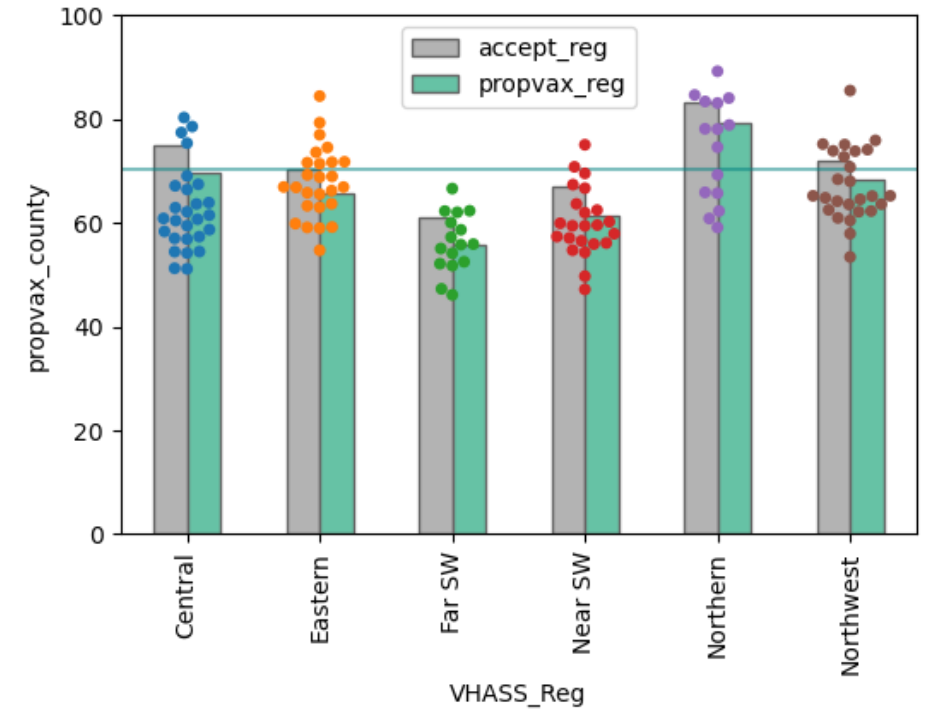


# Vaccination Acceptance by Region

## Corrections to surveys:

- Facebook administered survey is timely and broad, but biased by who accesses Facebook and answers the survey
- Correction approach:
  - Calculate an over-reporting fraction based on reported vaccinations compared to VDH administration data
  - Cross-validate coarse corrections against HPS survey at the state level and corrected in same manner

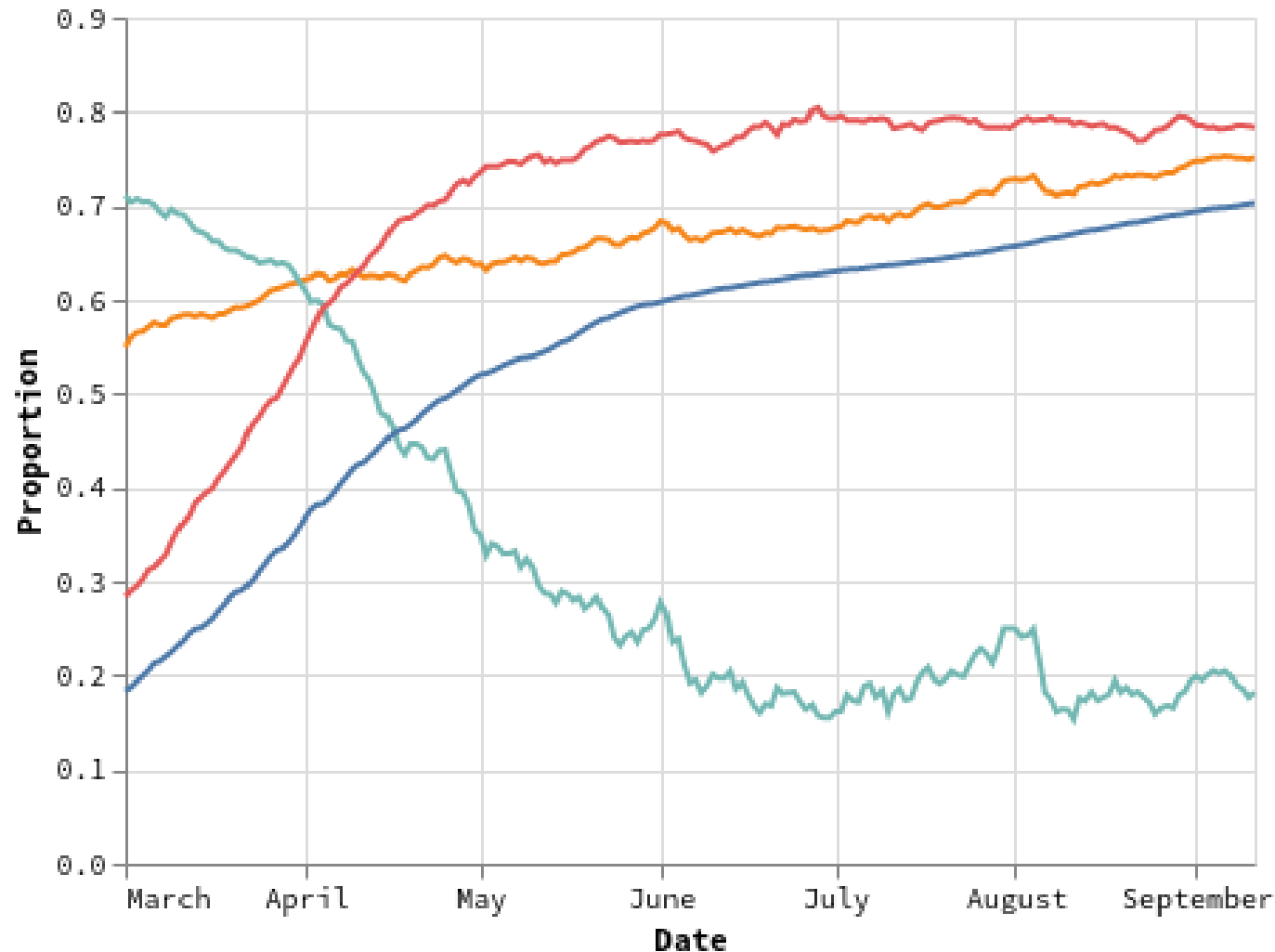
| Region          | COVIDcast accepting corrected | VDH proportion pop vaccinated |
|-----------------|-------------------------------|-------------------------------|
| Central         | 73%                           | 70%                           |
| Eastern         | 71%                           | 66%                           |
| Far SW          | 64%                           | 56%                           |
| Near SW         | 69%                           | 61%                           |
| Northern        | 84%                           | 79%                           |
| Northwest       | 72%                           | 68%                           |
| <b>Virginia</b> | <b>75%</b>                    | <b>70%</b>                    |



**Grey Bar:** Survey measured and corrected acceptance  
**Green Bar:** Proportion of eligible population administered a vaccine  
**Dots:** Proportion administered at least one dose for each county



# Vaccine Acceptance Components over Time



## Vaccine Willingness

- Administered Vaccines
- Corrected Acceptance
- Surveyed Vaccinated
- Unvaccinated Acceptance

## Vaccine Acceptance adjusted to include scheduled appointments

- Steady rise in acceptance over the past couple months
- Unvaccinated Acceptance shows ~20% of those who are unvaccinated are definitely or probably willing to be vaccinated
- Scheduled appointments for vaccination has increased through August but seems to be leveling off.

Data Source: <https://covidcast.cmu.edu>

16-Sep-21

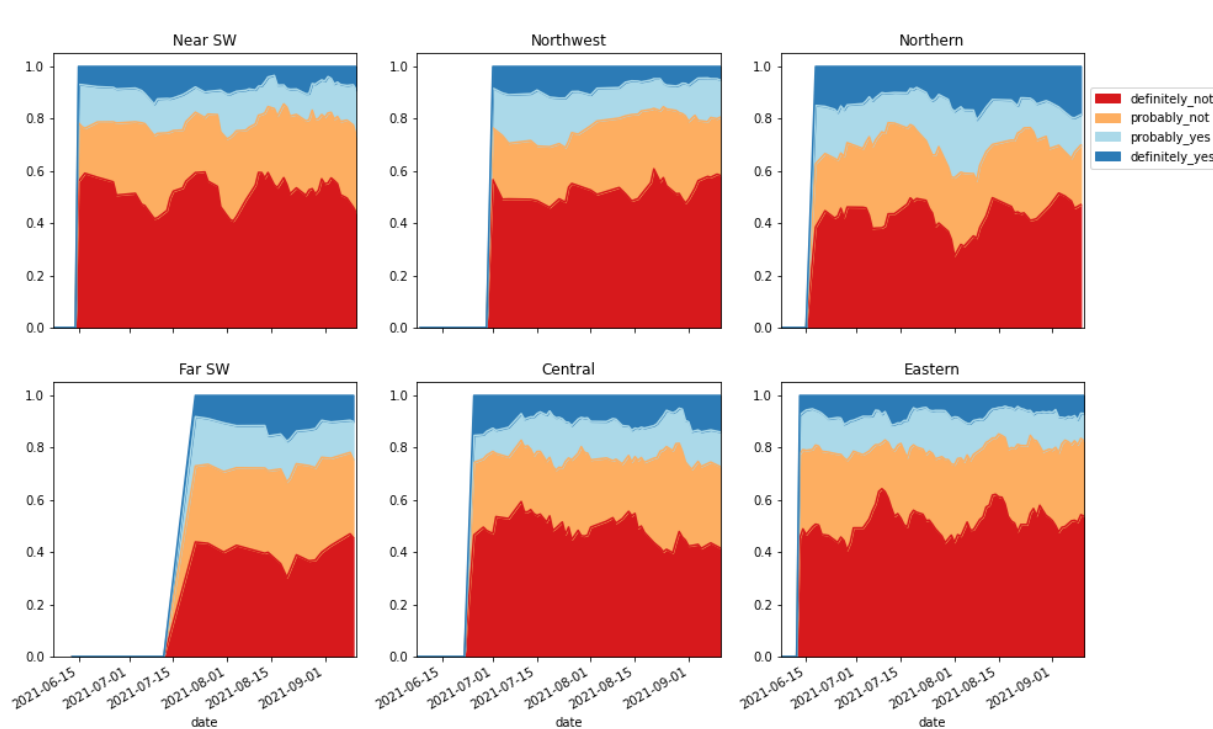


# Vaccine Acceptance by Region- COVIDcast

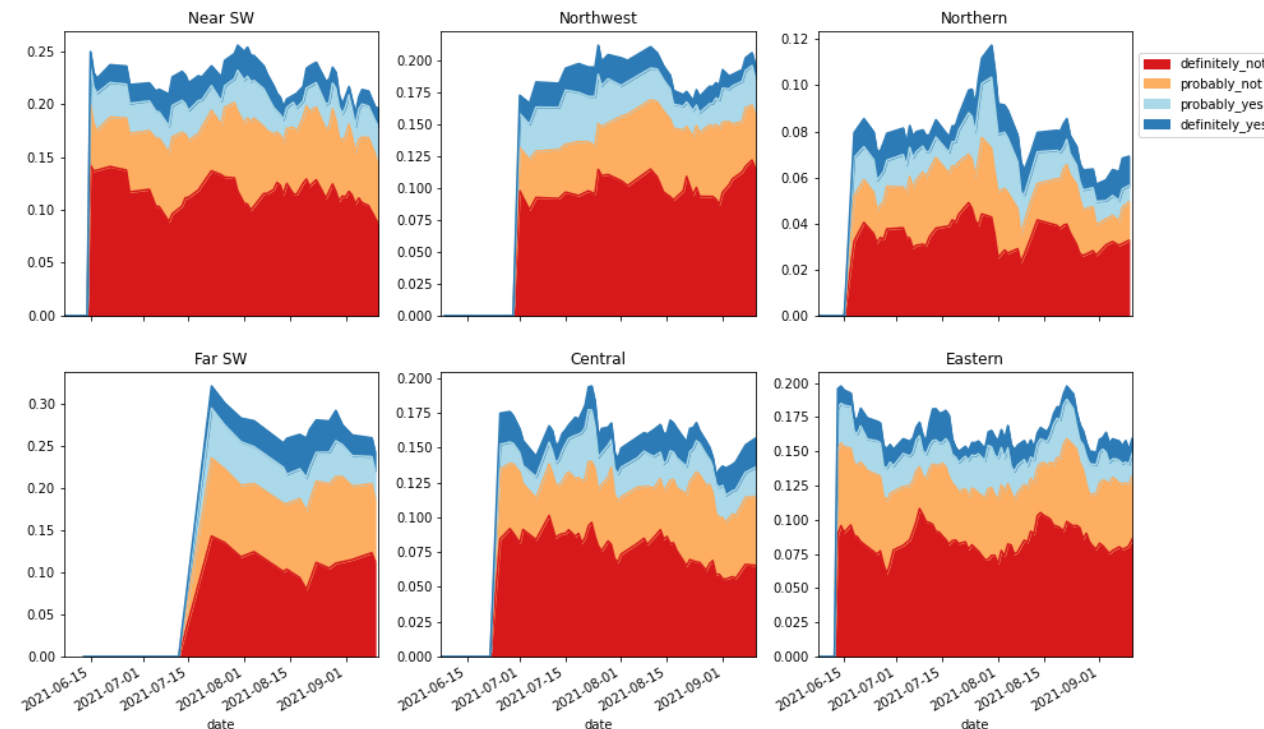
## Levels of Acceptance and potential acceptance in flux:

- Most regions (except Central and Far SW) see vaccine uptake in the “Definitely Yes”.
- Among the unvaccinated, about 20-30% remain in the Definitely/Probably “Yes” categories.
- About 50% of the Unvaccinated seem to be in the “Definitely Not” category.

### Unvaccinated Only



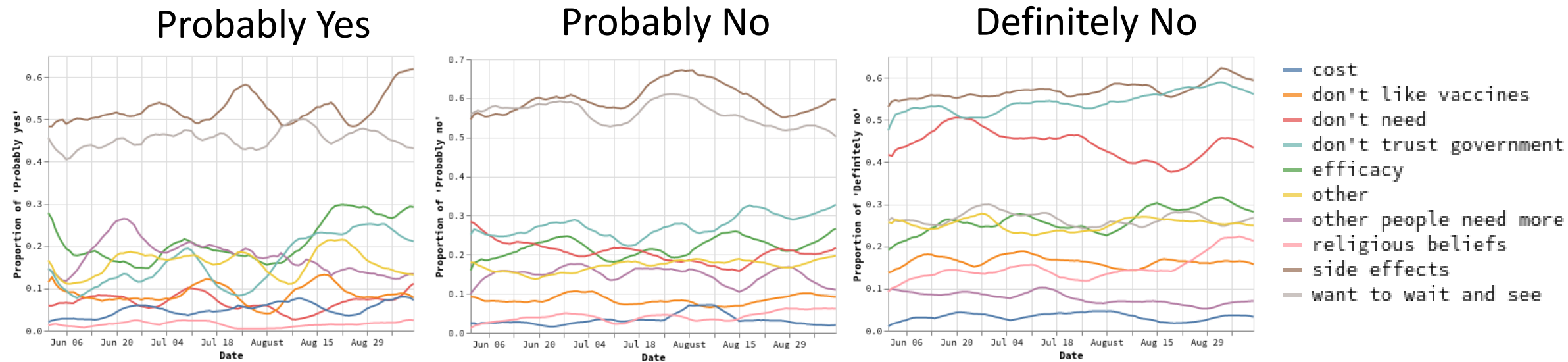
### All Respondents



Data Source: <https://covidcast.cmu.edu>

16-Sep-21

# Reasons for Hesitancy by Likelihood to Accept



## Reasons for Hesitancy vary across tiers of likelihood to accept the vaccine

- Probably Yes and Probably No most concerned about side effects & are waiting to see
- Definitely No are concerned about side effects but also don't think they need the vaccine and don't trust the government, though don't need is declining
- Most other reasons are below 30% within these tiers of likelihood

# Mask Usage Increases

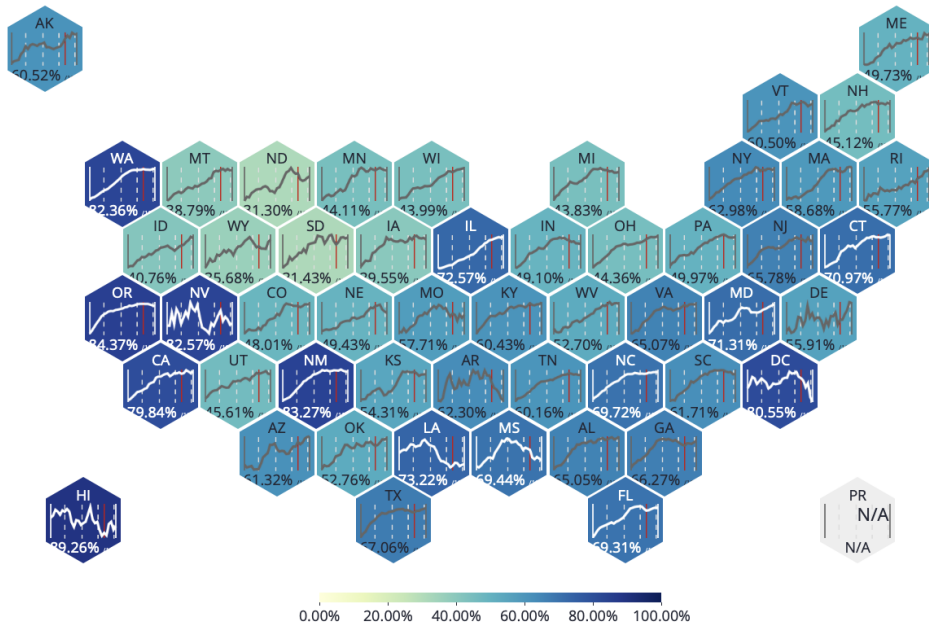
## Self-reported mask usage has declined for months, but rebounded

- State-wide continues to rise, now outpaces US (65% vs. 64%)
- Progress in some counties has stalled or declined

### PEOPLE WEARING MASKS MAP

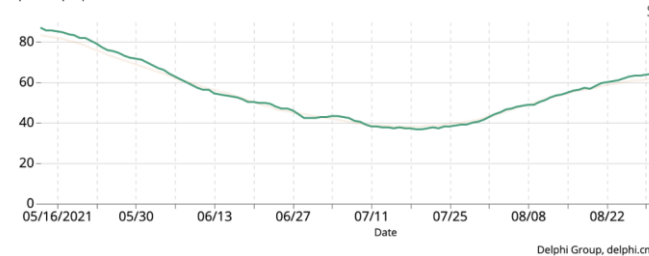
Click on a state to show this region

Show US States as Beehive Grid ☒ Show US Counties as Choropleth Map



### PEOPLE WEARING MASKS CHART

People Wearing Masks in Virginia  
per 100 people



☒ Rescale Y-axis ☐ Show All Dates

● Virginia

65.22% per 100

● United States

63.81% per 100

### VIRGINIA COUNTIES

| COUNTY                    | CHANGE LAST 7 DAYS | PER 100     | HISTORICAL TREND<br>8/16 9/13 |
|---------------------------|--------------------|-------------|-------------------------------|
| United States             | → +1.58%           | 63.62% /100 |                               |
| Virginia                  | → +1.95%           | 65.07% /100 |                               |
| Newport News, VA          | ↓ -18.65%          | 61.49% /100 |                               |
| Virginia Beach, VA        | → -0.32%           | 61.57% /100 |                               |
| Norfolk, VA               | ↑ +7.82%           | 65.89% /100 |                               |
| Chesapeake, VA            | ↑ +9.63%           | 67.46% /100 |                               |
| Albemarle County, VA      | ↑ +9.42%           | 70.03% /100 |                               |
| Chesterfield County, VA   | ↑ +9.27%           | 70.03% /100 |                               |
| Henrico County, VA        | → +2.68%           | 72.22% /100 |                               |
| Richmond, VA              | → +1.69%           | 73.31% /100 |                               |
| Prince William County, VA | ↑ +5.16%           | 75.51% /100 |                               |
| Loudoun County, VA        | → +5.07%           | 76.46% /100 |                               |
| Hampton, VA               | ↑ +12.24%          | 77.00% /100 |                               |
| Fairfax, VA               | → +2.52%           | 77.89% /100 |                               |
| Arlington County, VA      | ↑ +46.84%          | 79.97% /100 |                               |

Data Source: <https://covidcast.cmu.edu>  
16-Sep-21



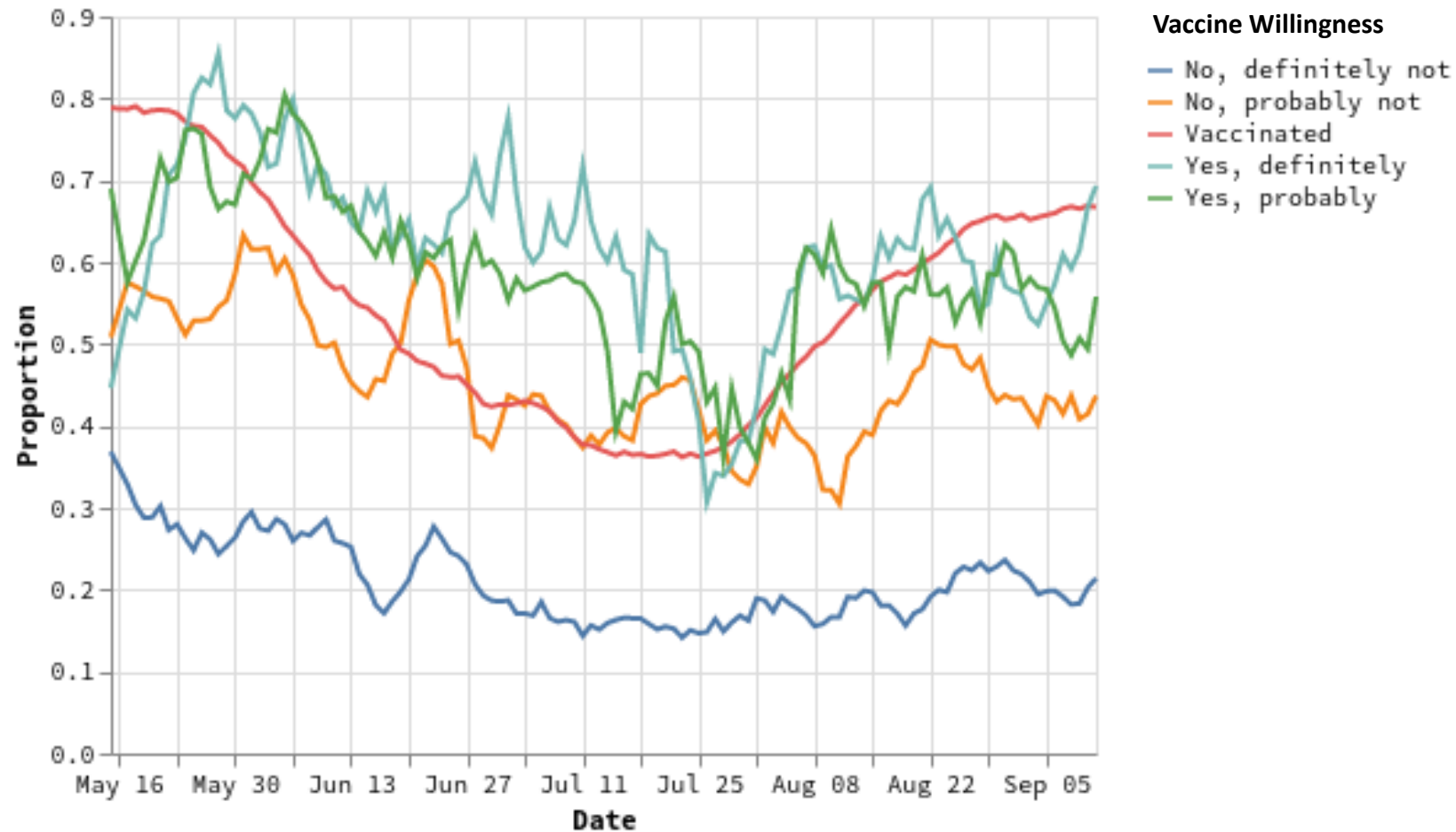
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# Mask Wearing by Vaccine Willingness

**Among the different tiers of vaccine acceptance, mask wearing increasing**

- Only those who would “definitely not” take the vaccine if offered have a low level of mask usage
- Probably Yes, joins vaccinated with highest mask wearing at ~67%



Data Source: <https://covidcast.cmu.edu>

16-Sep-21

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# SARS-CoV2 Variants of Concern

Emerging new variants will alter the future trajectories of pandemic and have implications for future control

- Emerging variants can:
  - Increase transmissibility
  - Increase severity (more hospitalizations and/or deaths)
  - Limit immunity provided by prior infection and vaccinations
- Genomic surveillance remains very limited
  - Challenges ability to estimate impact in US to date and estimation of arrival and potential impact in future

|           | New WHO Name | Transmissibility | Immune Evasiveness | Vaccine Effectiveness <sup>^</sup> |
|-----------|--------------|------------------|--------------------|------------------------------------|
| Ancestral |              | —                | —                  | ✓                                  |
| D614G     |              | +                | —                  | ✓                                  |
| B.1.1.7   | Alpha        | +++              | —                  | ✓                                  |
| B.1.351   | Beta         | +                | ++++               | ✓                                  |
| P.1       | Gamma        | ++               | ++                 | ✓                                  |
| B.1.429   | Epsilon      | +                | +                  | ✓                                  |
| B.1.526   | Iota         | +                | +                  | ✓                                  |
| B.1.617.2 | Delta        | ++++*            | ++ <sup>#</sup>    | ✓                                  |

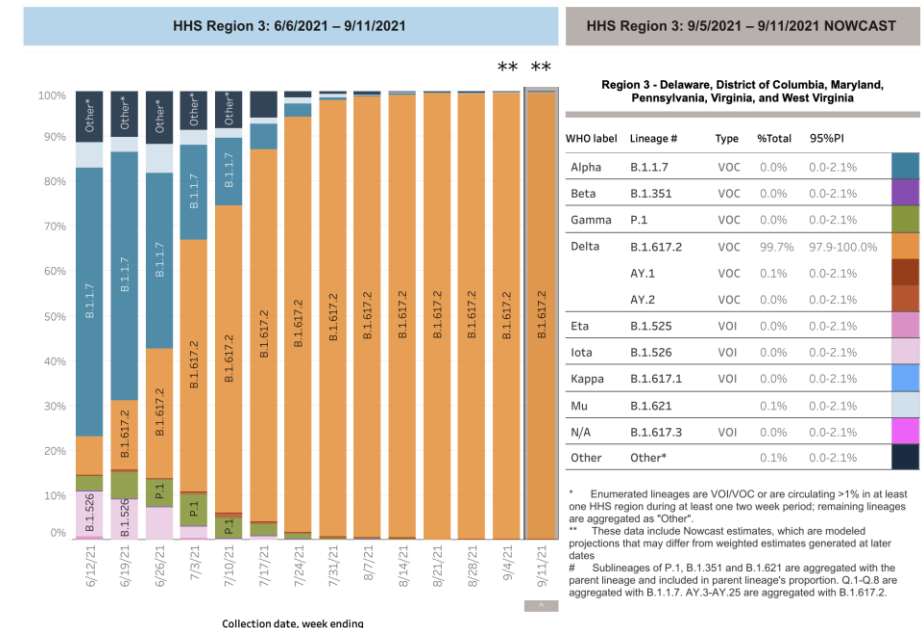
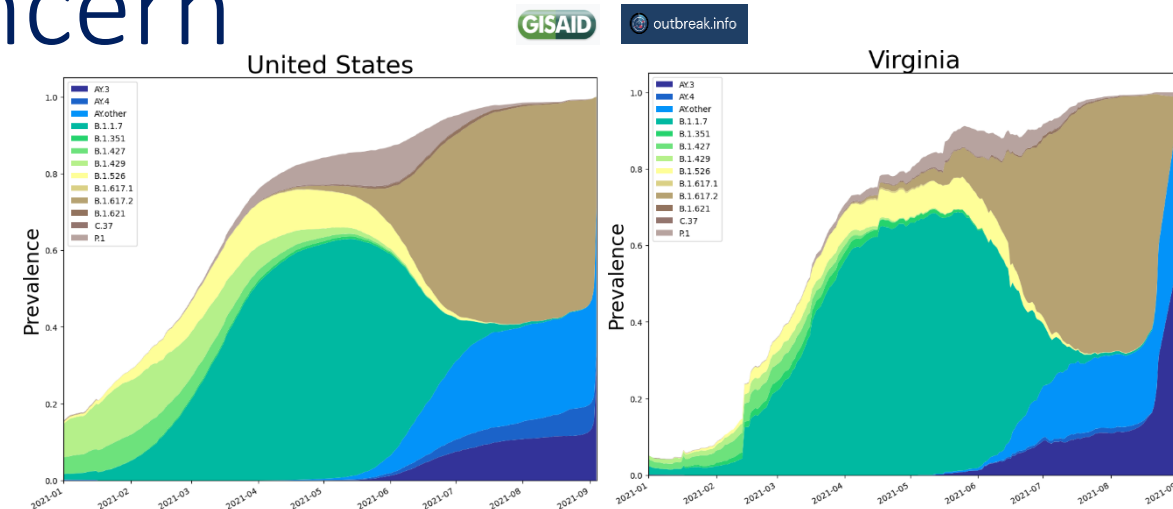
<sup>\*</sup>Relative transmissibility to B.1.1.7 yet to be fully defined

<sup>^</sup>Effectiveness from real world evidence vs. severe illness, not all vaccines are effective vs all variants, and importance of 2-doses, especially for B.1.617.2 for which 1 dose of mRNA or AZ is only ~30% effective <sup>#</sup> May carry more immune escape than P.1, to be determined



World Health Organization

WHO and Eric Topol

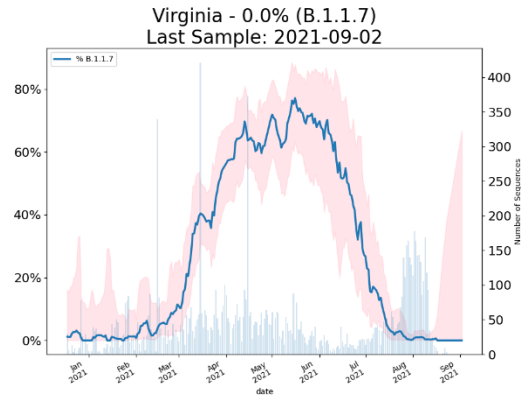




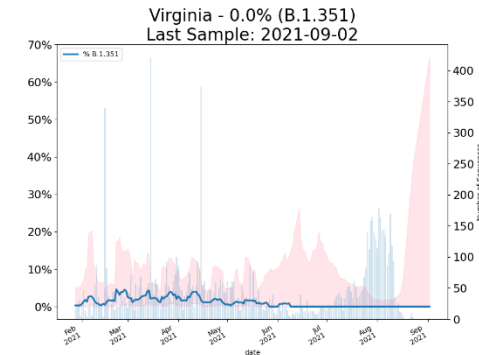
# SARS-CoV2 Variants of Concern

## Previous Variants

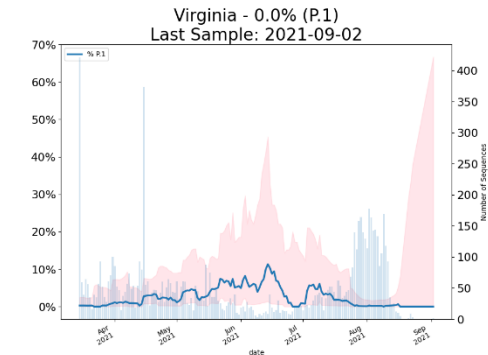
### Alpha $\alpha$ - Lineage B.1.1.7



### Beta $\beta$ - Lineage B.1.351

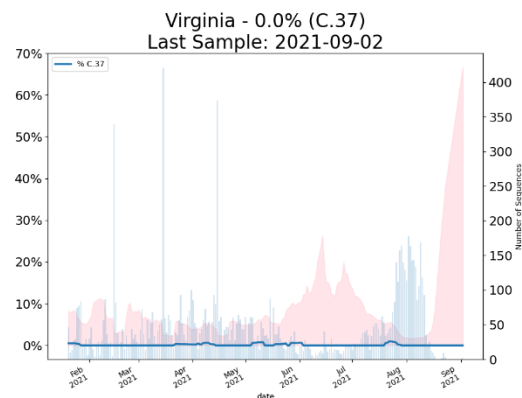


### Gamma $\gamma$ - Lineage P.1

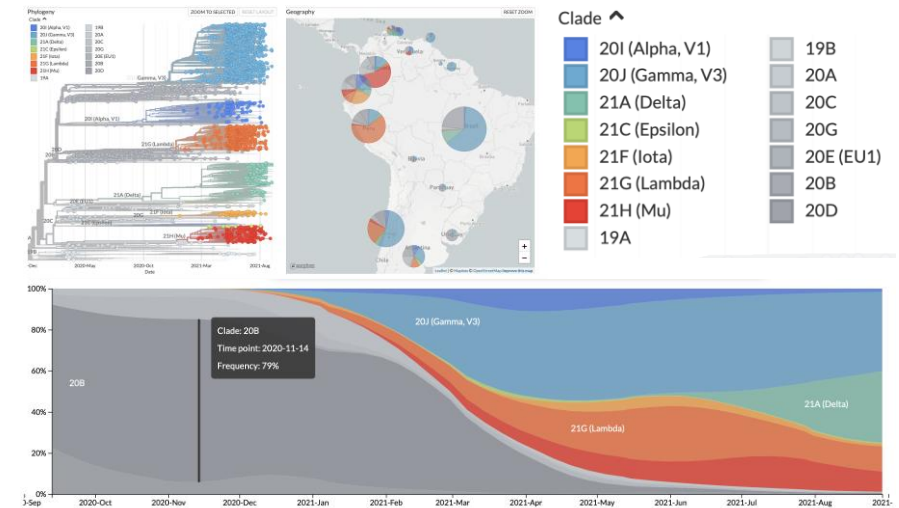
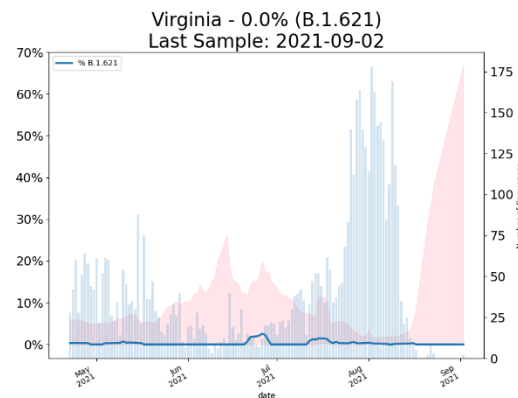


## Emerging Variants

### Lambda $\lambda$ - Lineage C.37



### Mu $\mu$ - Lineage B.1.621

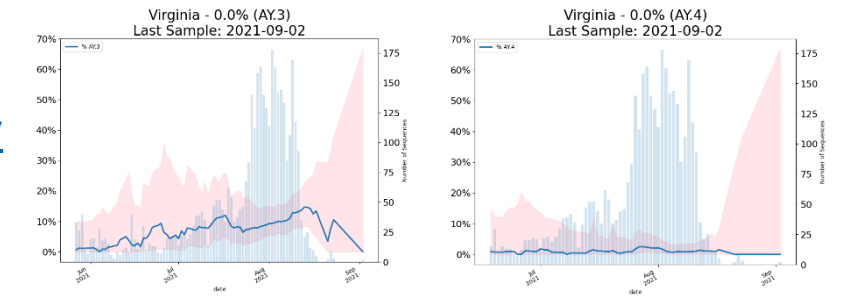
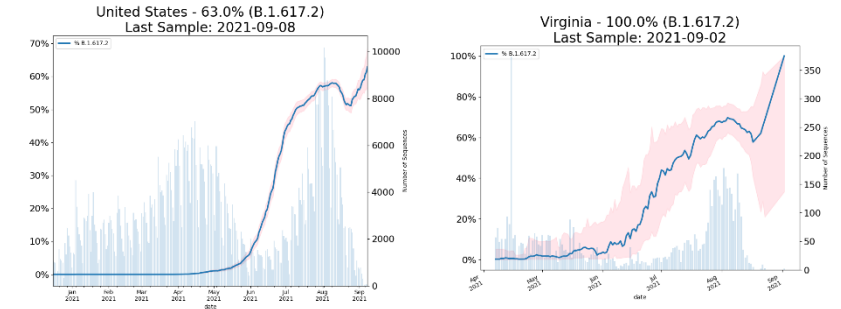


Delta continues to out compete Lambda and Mu in South America  
[Trevor Bedford Tweet](#) & [Nextstrain Analysis](#)

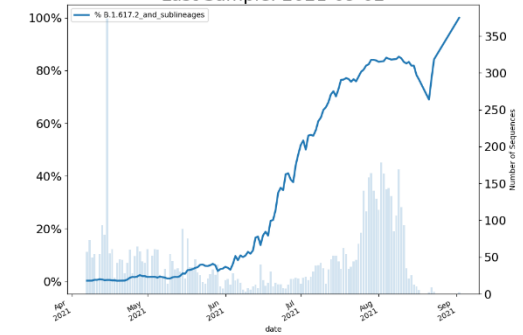
# SARS-CoV2 Variants of Concern

## Delta $\delta$ - Lineage B.1.617.2 and related subvariants

- Delta plus  $\delta+$  lineage which contains the K417N mutation is emerging as a sub-variant that is even more transmissible; declared a VoC in India
- Delta variant now dominates most of Europe and US
- CDC recommends resumption of mask wearing indoors due to reports of breakthrough infections of the vaccinated possibly being transmissible
- [Recent study from Mayo clinic](#) shows Delta reducing the efficacy of mRNA vaccines (Pfizer more so than Moderna) along with [other reports](#). [Israeli study](#) showed 64% efficacy against infection, however, a 3<sup>rd</sup> dose may [counteract this reduction](#)
- [Public Health Scotland study in Lancet](#) suggests Delta is 2x more likely to cause hospitalization than Alpha
- Subvariants AY.3 (15%) and AY.4 (1.5%) of Delta are more prevalent, these subvariants are mainly clustered in the US, others mainly outside of US



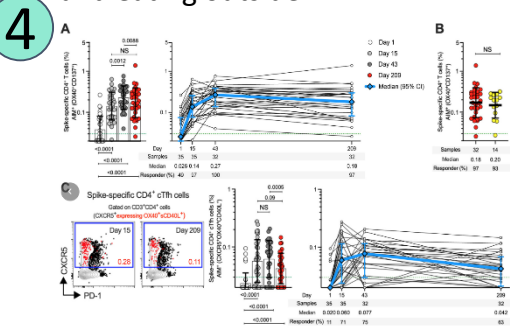
Virginia - 100.0% ([B.1.617.2', 'AY.1', 'AY.10', 'AY.11', 'AY.12', 'AY.2', 'AY.3', 'AY.3.1', 'AY.4', 'AY.5', 'AY.6', 'AY.7', 'AY.9



Main delta and all other subvariants

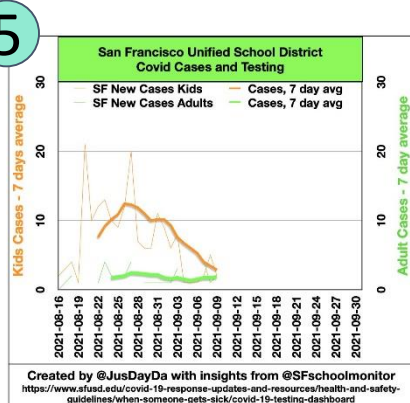
# Variants & Vaccines

1. [Public Health England report](#) verifies strong protection against hospitalization and highlights that longer intervals between vaccine doses can be more efficacious, but too long may also diminish efficacy
2. Additional analyses on the PHE study above demonstrate the utility of additional doses for the most vulnerable populations, and benefits of dose gaps larger than 4 weeks.
3. [Updated analysis](#) on Israeli data that avoids Simpson's paradox continues to demonstrate vaccines high effectiveness against severe disease, and illustrates that lumping the under 12 population (ineligible for vaccination) into analyses can further diminish measured effectiveness
4. Study in [Science](#) illustrates the equivalency of low-dose Moderna mRNA vaccine with natural immunity, including longer term T-cell responses, suggesting potential for dose sharing (25µg vs. 100 µg)
5. San Francisco schools limited outbreaks and found very limited number of cases in their schools through basic low-cost infection control measures: vaccines, ventilation. HEPA filter. monitor CO2. indoor masks. and eating outside



Moderna vaccine stimulated cellular immunity vaccine-generated spike-specific memory CD4+ T cells 6 months post-boost were comparable in quantity and quality to COVID-19 cases. Spike-specific CD8+ T cells were generated in 88% of subjects, with equivalent memory at 6 months post-boost compared to COVID-19 cases.

<https://www.science.org/doi/10.1126/science.abj9853>



Everything is trending in the right direction at SF schools.  
 #StaySteady  
 #Vaccinate  
 #Ventilate fan in window blowing out  
 #useHEPA to filter Covid aerosols 4-6 air changes per hour  
 #monitorCO2 under 800ppm=breathe less air from others lungs  
 #MaskUpN95 indoors  
 #EatOutdoors

Twitter

## 3 Israel Severe Cases Aug10-Sept8 (Age<12 pulled out)

| Age      | Population (%)       |                      |                      | Total Serious Cases (per 100k) |                  |                 | VE <sub>SERIOUS</sub><br>(% reduction relative to unvaccinated) |       |
|----------|----------------------|----------------------|----------------------|--------------------------------|------------------|-----------------|---|-------|
|          | Not Vax              | Vax (NB)             | Boost                | Not Vax                        | Vax (NB)         | Boost           | Vax (NB)  | Boost |
| All ages | 1,226,932<br>(17.8%) | 4,446,814<br>(64.7%) | 1,200,310<br>(17.5%) | 9289<br>(757.1)                | 7688<br>(172.9)  | 1378<br>(114.8) | 77.2%   | 84.8% |
| >12      | 2,229,128            | 0                    | 0                    | 0                              | 0                | 0               | *   | *     |
| 12-60    | 1,071,666<br>(20.1%) | 3,870,427<br>(72.6%) | 385,137<br>(7.2%)    | 3716<br>(346.75)               | 1437<br>(37.1)   | 100<br>(26.0)   | 89.3%   | 92.5% |
| >60      | 157,273<br>(10.2%)   | 576,344<br>(37.2%)   | 815,106<br>(52.6%)   | 5573<br>(3543.5)               | 6251<br>(1084.6) | 1278<br>(156.8) | 69.4%   | 95.6% |

- These data include all children <12yr in these data, even though they are not eligible for vaccination. Given children are at a much lower risk of severe COVID-19, this strongly attenuates VE estimates for the <60 age group and overall, with Age<12yr as an exceptionally strong Simpson's Effect factor
- Comparing ministry tables with and without <12yr, we can estimate this group as 2,229,128.
- Assuming zero serious COVID-19 & zero vaccinations from Aug10-Sept8, the numbers change substantially.

## Israel Severe Cases Aug10-Sept8

From: <https://datadashboard.health.gov.il/COVID-19/general>

| Age      | Population (%)    |                   |                   | Total Serious Cases (per 100k) |               |              | VE <sub>SERIOUS</sub> (% reduction relative to unvaccinated) |       |
|----------|-------------------|-------------------|-------------------|--------------------------------|---------------|--------------|--|-------|
|          | Not Vax           | Vax (NB)          | Boost             | Not Vax                        | Vax (NB)      | Boost        | Vax (NB)   | Boost |
| All ages | 3,456,060 (38.0%) | 4,446,814 (48.8%) | 1,200,310 (13.2%) | 9289 (268.6)                   | 7688 (172.9)  | 1378 (114.8) | 35.6%  | 57.3% |
| <60      | 3,300,794 (43.7%) | 3,870,427 (51.2%) | 385,137 (5.1%)    | 3716 (112.6)                   | 1437 (37.1)   | 100 (26.0)   | 67.0%  | 76.9% |
| >60      | 157,273 (10.2%)   | 576,344 (37.2%)   | 815,106 (52.6%)   | 5573 (3543.5)                  | 6251 (1084.6) | 1278 (156.8) | 69.4%  | 95.6% |

- The MOH data doesn't have exact counts for # vaccinated/not vaccinated/boosted for each day.
- However, they have raw counts and normalized counts per 100k for each outcome (vaccines, serious cases, deaths) from which it is possible to estimate the exact counts arithmetically. We used vaccination numbers. These numbers change daily so here, we average over the inferred number vaccinated/not vaccinated/boosted across days.
- The numbers don't perfectly add up because of small roundoff errors in posted data, but are close enough to not substantially affect VE numbers

<https://www.covid-datascience.com/post/what-do-new-israeli-data-say-about-effect-of-vaccines-boosters-vs-death-critical-severe-disease>

1

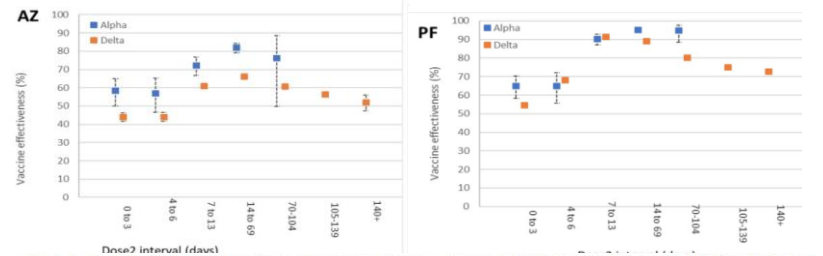
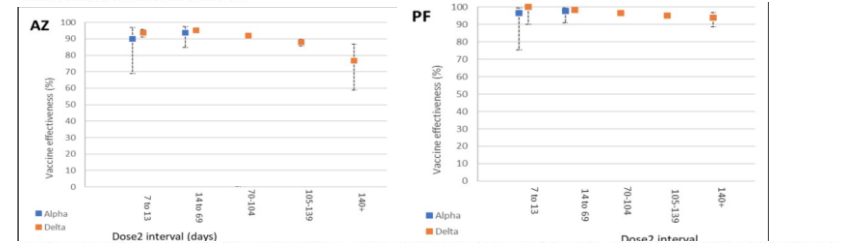


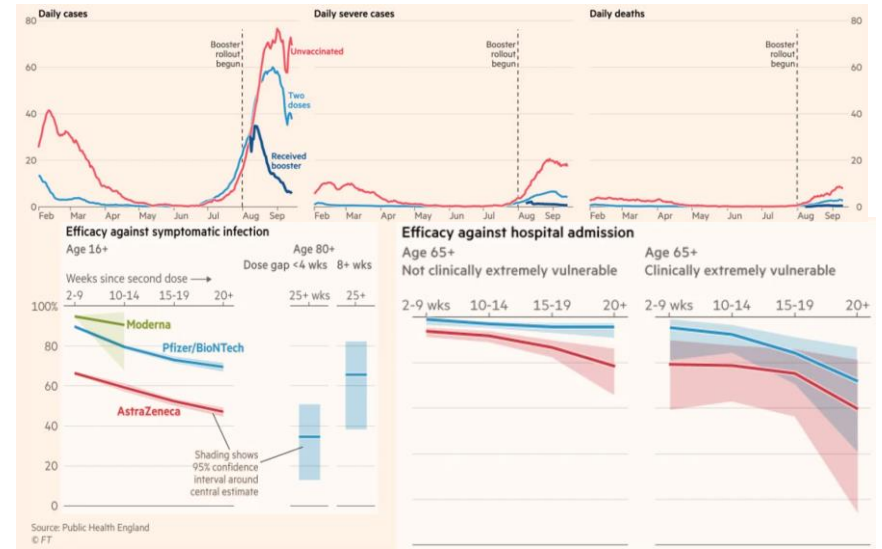
Figure 1: vaccine effectiveness against symptomatic disease - all ages (a) AstraZeneca Vaxzevria, (b) Pfizer-BioNTech Comirnaty, (c) Moderna Spikevax



vaccine effectiveness against hospitalisation - all ages (a) AstraZeneca Vaxzevria, (b) Pfizer-BioNTech Comirnaty

In a recent report PHE conducted a test negative case control design study to estimate VE against symptomatic disease, hospitalisation and death. They compared vaccination status in persons with symptomatic Covid-19 with vaccination status in persons who reported symptoms but had a negative test. This approach helps to control for biases related to health-seeking behaviour, access to testing, and case ascertainment. [PHE report](#)

2

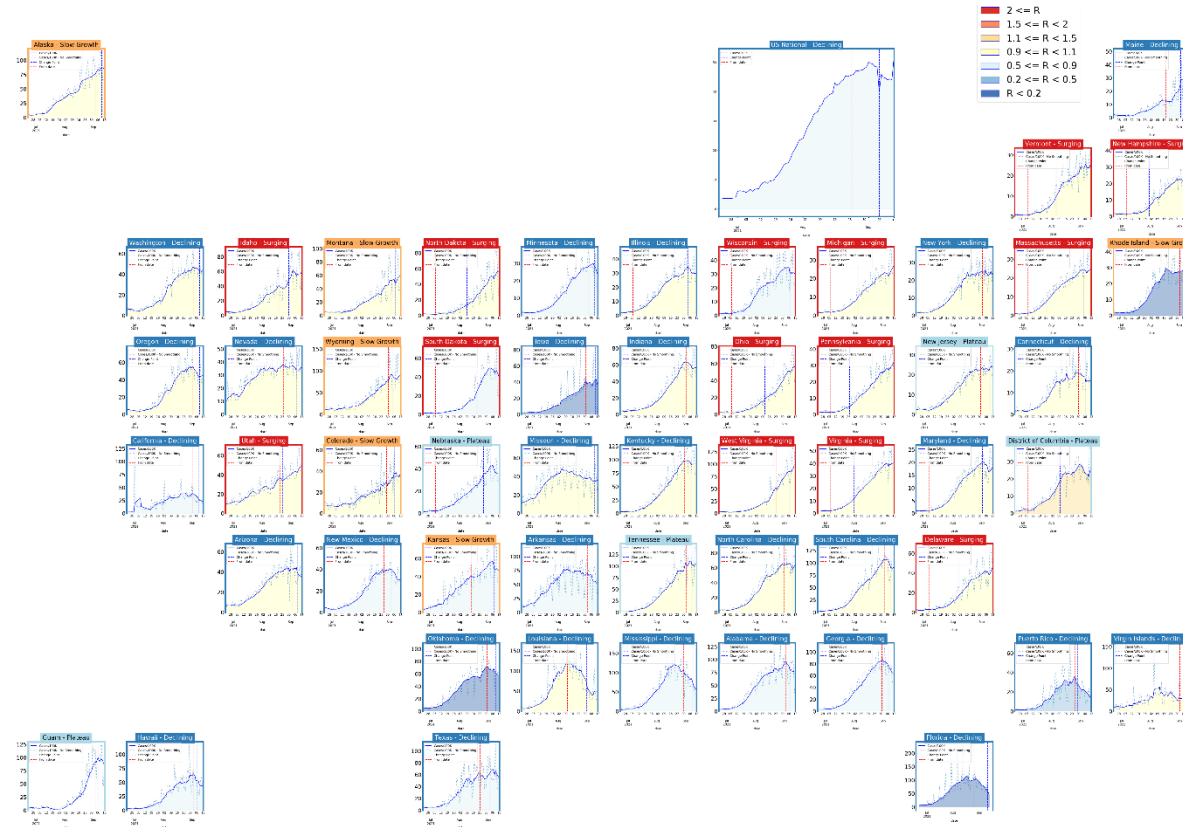


A recent Financial Times article based on data from PHE raises important aspects in the case for boosters, namely 1) the dosage gap impact on immune resilience and 2) the likelihood of being infected based on waning immunity, age, and other aspects of vulnerability. <https://www.ft.com/content/cf83b3a1-fe06-4cf9-999c-7500090aee7c>



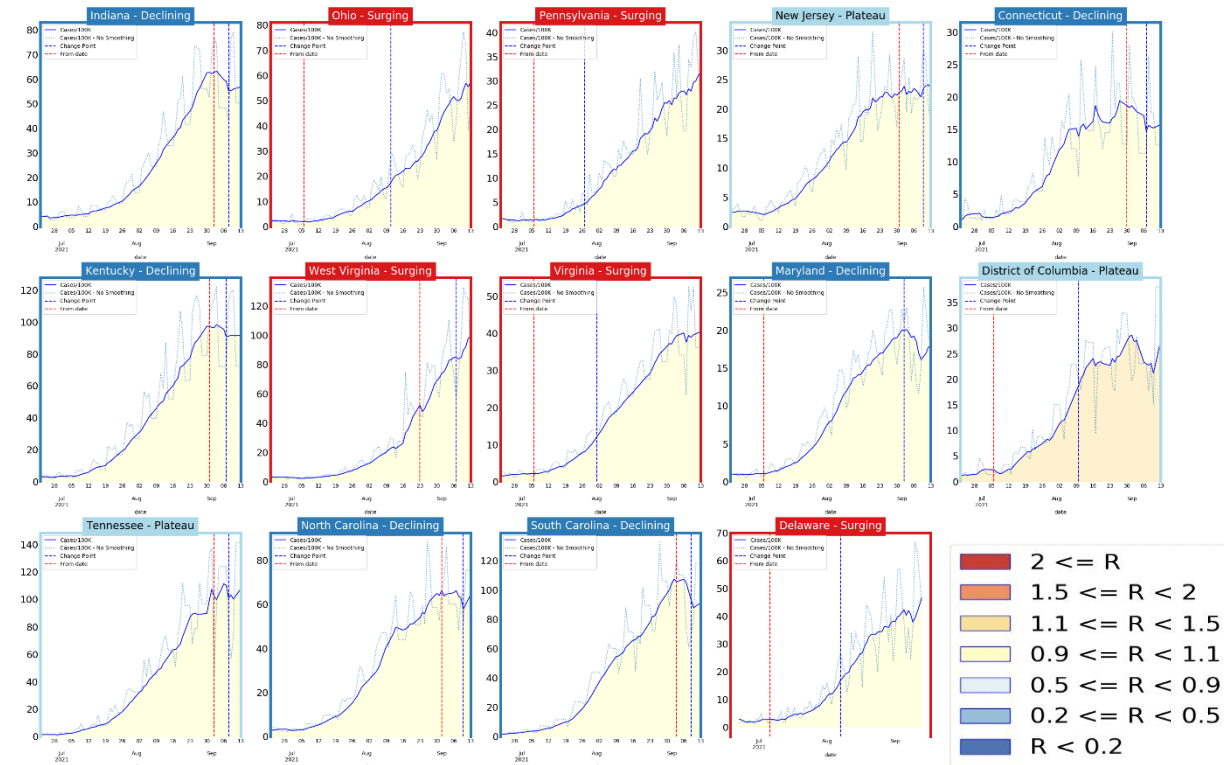
# Other State Comparisons

## Trajectories of States



- More of the country has plateaued and started to decline
- Many states remain in surge, but show signs of slowing
- Case rates remain very high, but nationally rates are starting to come down

## Virginia and her neighbors

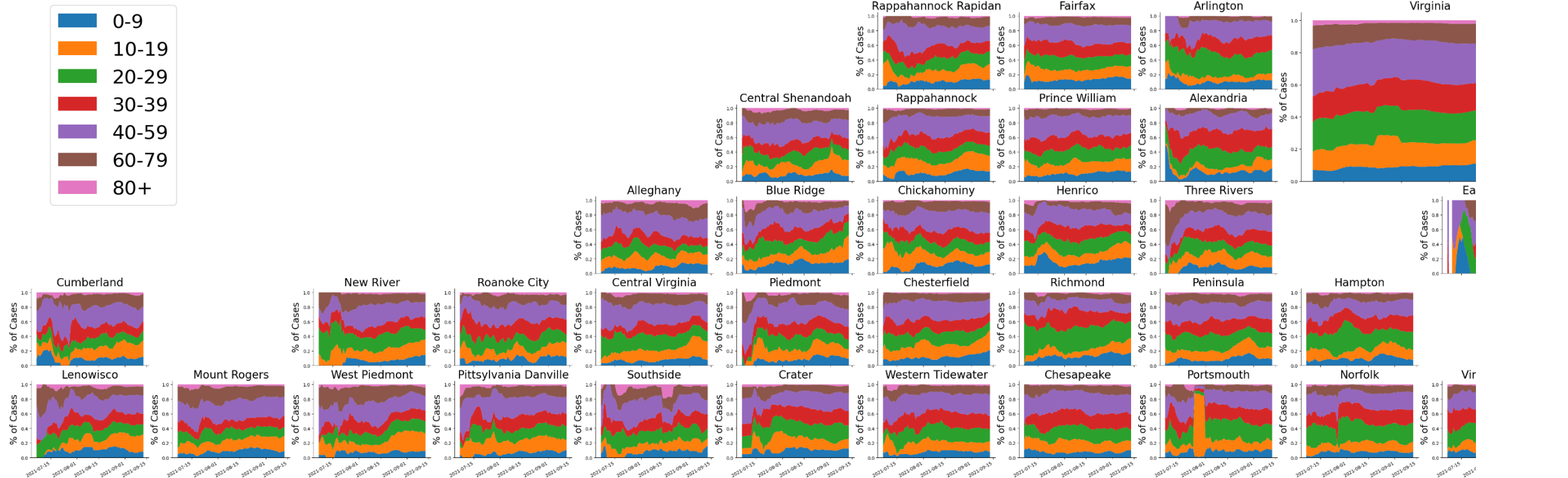


- Nearly all states show signs of plateau or slowing growth, with several declining in the past week
- Case rates remain high

# Age-Specific Case Rates

## Case Rates (per 100K) by Age Groups

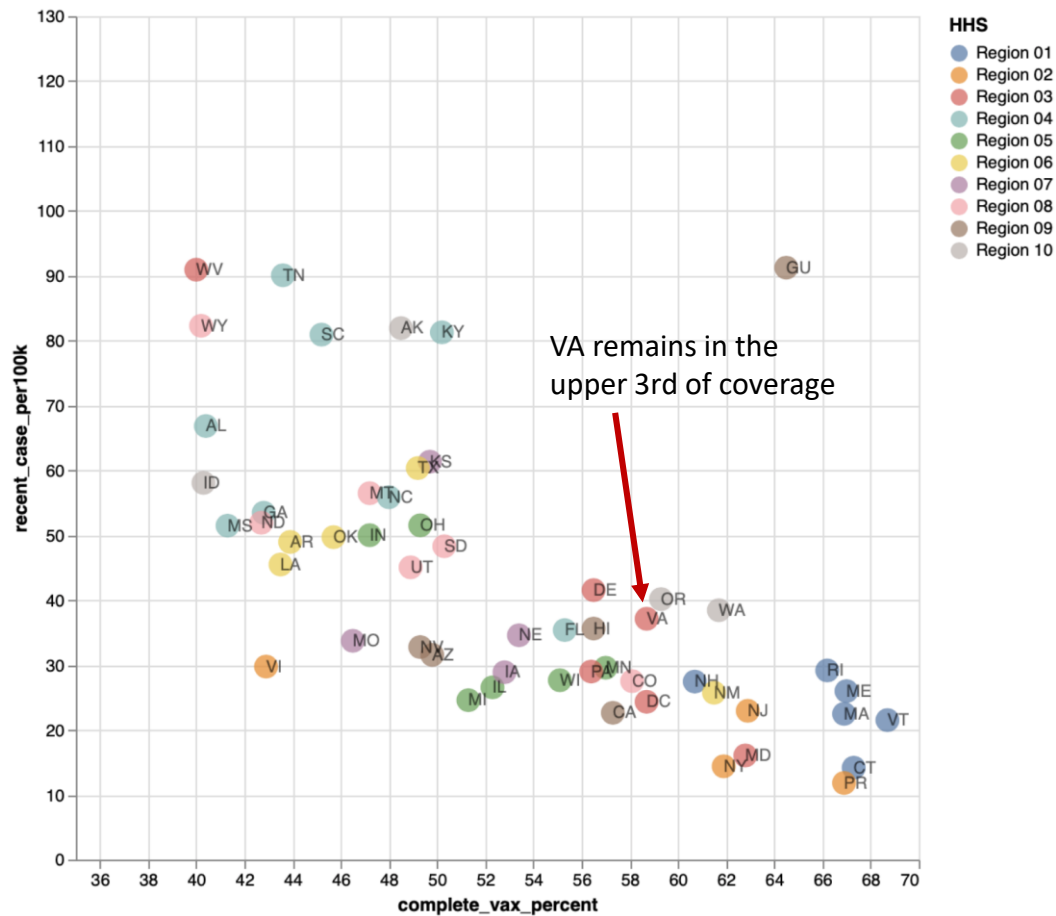
- Rapid growth in many regions in the 0-19 age range, many exceeding 20% of current case rates (nearly 10% for 0-9)
- Case Rate in under 40 group far exceeds 40+ in all districts



# Recent Cases Correlate with Vax Coverage

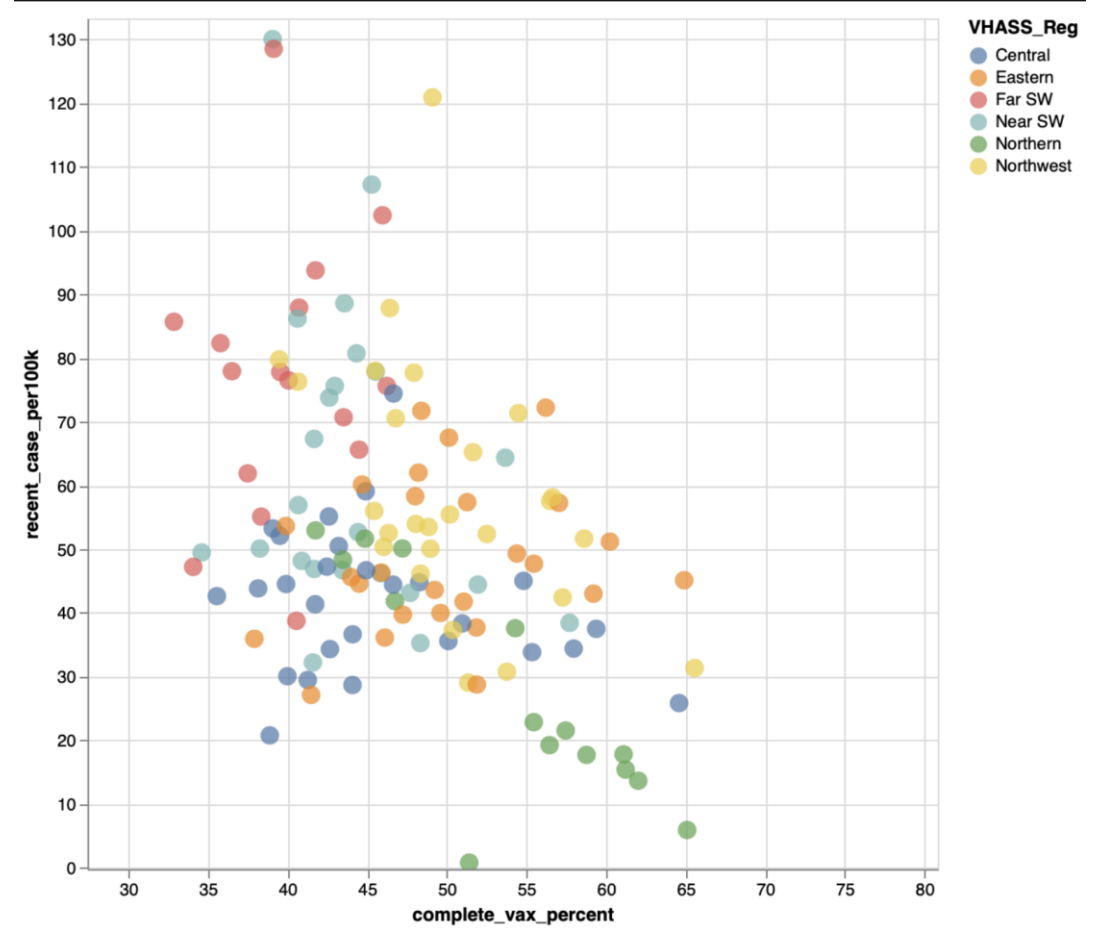
## Mean cases per 100K vs. vaccine coverage

- States with lower vax coverage have had the worst case spikes



## Virginia Counties

- Counties with higher vax coverage are maintaining lower case rates



# Zip code level weekly Case Rate (per 100K)

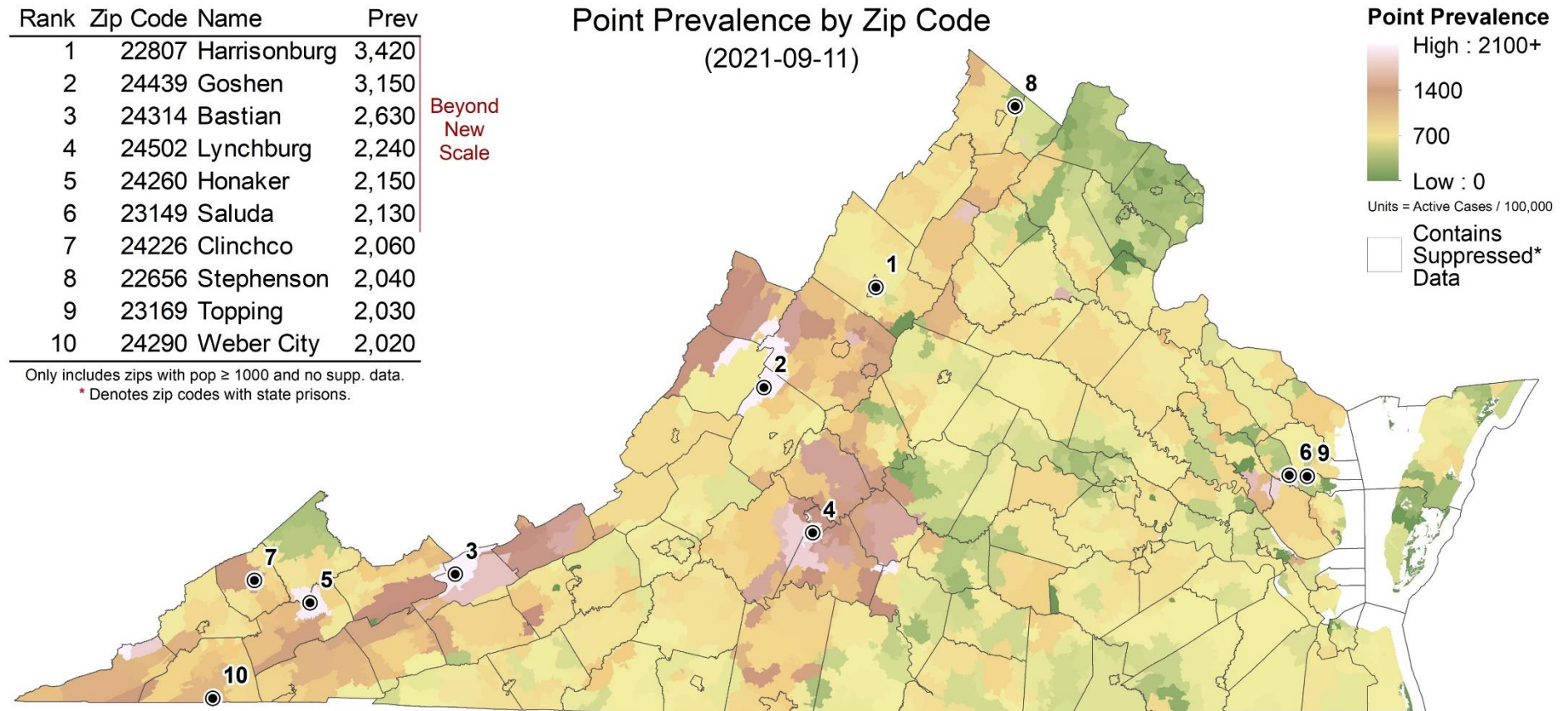
## Case Rates in the last week by zip code

- Color scaled adjusted to accommodate the very high prevalence levels this week
- Clusters of high prevalence in Southwest and Eastern
- Some counts are low and suppressed to protect anonymity, those are shown in white

| Rank | Zip Code | Name         | Prev  |
|------|----------|--------------|-------|
| 1    | 22807    | Harrisonburg | 3,420 |
| 2    | 24439    | Goshen       | 3,150 |
| 3    | 24314    | Bastian      | 2,630 |
| 4    | 24502    | Lynchburg    | 2,240 |
| 5    | 24260    | Honaker      | 2,150 |
| 6    | 23149    | Saluda       | 2,130 |
| 7    | 24226    | Clinchco     | 2,060 |
| 8    | 22656    | Stephenson   | 2,040 |
| 9    | 23169    | Topping      | 2,030 |
| 10   | 24290    | Weber City   | 2,020 |

Only includes zips with pop  $\geq 1000$  and no supp. data.

\* Denotes zip codes with state prisons.



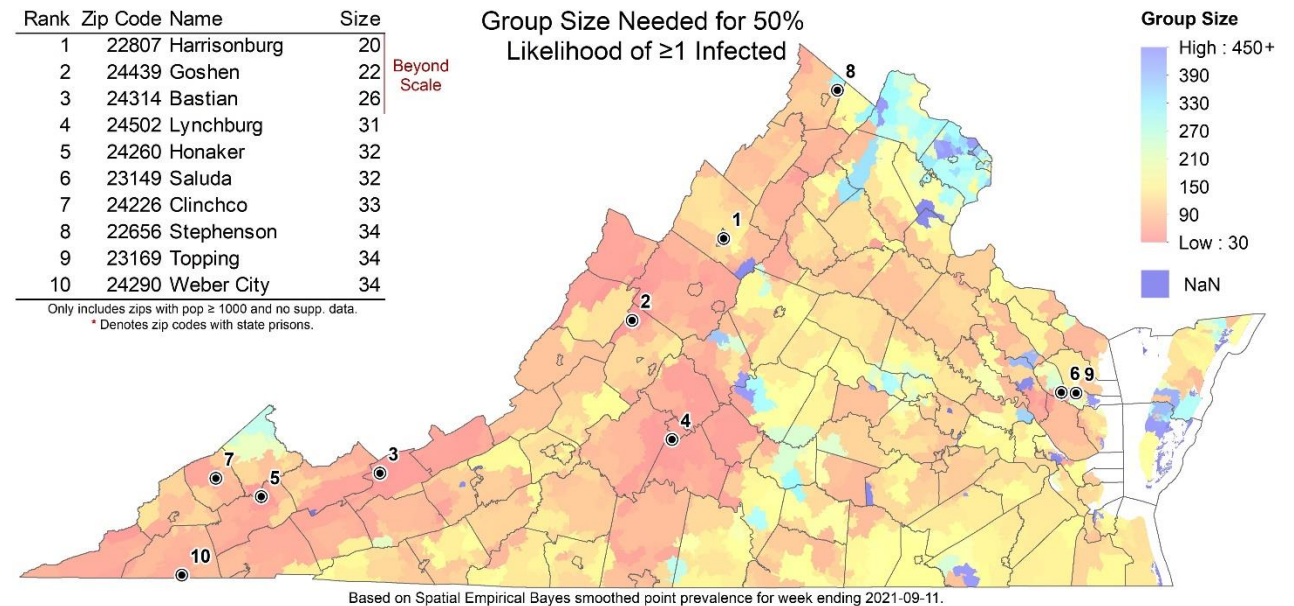
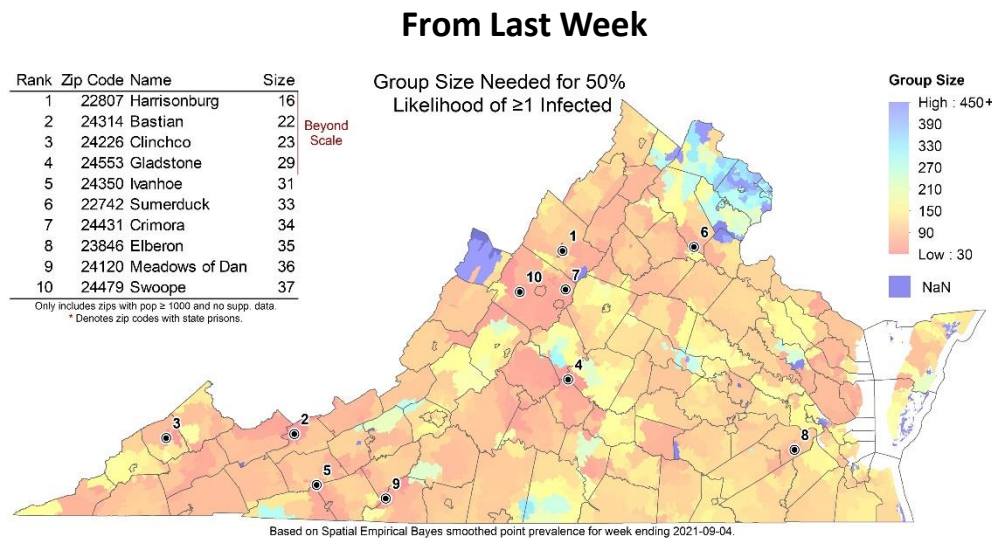
Based on Spatial Empirical Bayes smoothed point prevalence for week ending 2021-09-11.



# Risk of Exposure by Group Size

**Case Prevalence in the last week by zip code used to calculate risk of encountering someone infected in a gathering of randomly selected people (group size 25)**

- **Group Size:** Assumes 2 undetected infections per confirmed case (ascertainment rate from recent seroprevalence survey), and shows minimum size of a group with a 50% chance an individual is infected by zip code (eg in a group of 20 in Harrisonburg, there is a 50% chance someone will be infected)



# HCW Prevalence

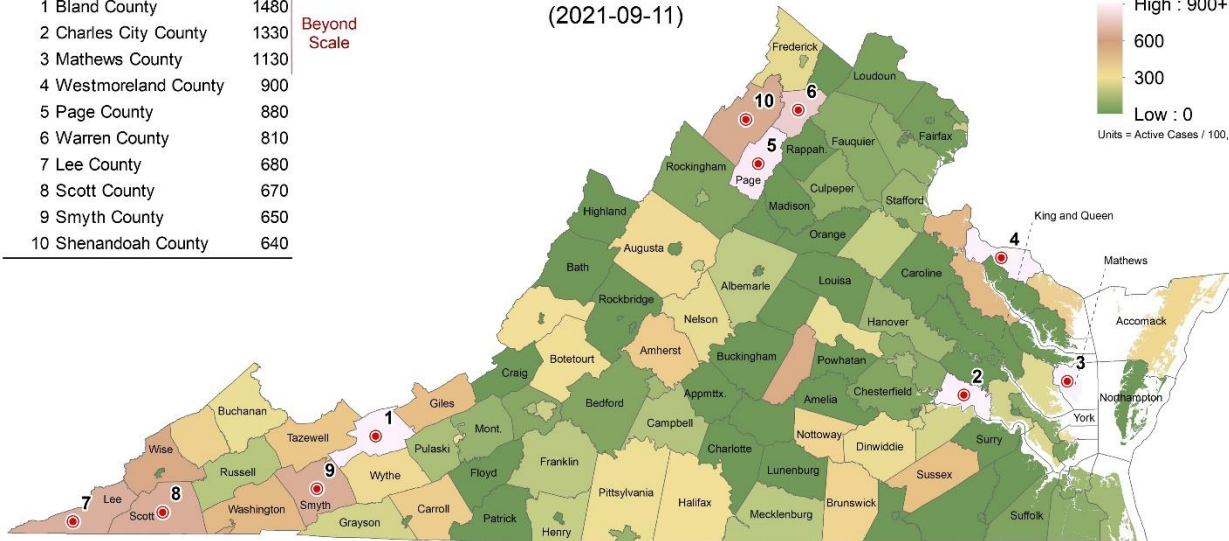
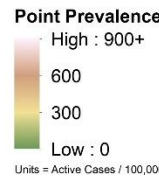
- **HCW prevalence:** Case rate among health care workers (HCW) in the last week using patient facing health care workers as the denominator
  - Clusters of high HCW point prevalence in far southwest (Wise & Dickinson Counties) and south of Richmond (Lunenburg and Prince Edward to Surry Counties)
- **HCW Ratio:** HCW Prevalence / Total Case Prevalence
  - (blue = higher case rate among public, red = higher case rate among HCW)

## HCW Prevalence

| Rank | Name                | Prev |
|------|---------------------|------|
| 1    | Bland County        | 1480 |
| 2    | Charles City County | 1330 |
| 3    | Mathews County      | 1130 |
| 4    | Westmoreland County | 900  |
| 5    | Page County         | 880  |
| 6    | Warren County       | 810  |
| 7    | Lee County          | 680  |
| 8    | Scott County        | 670  |
| 9    | Smyth County        | 650  |
| 10   | Shenandoah County   | 640  |

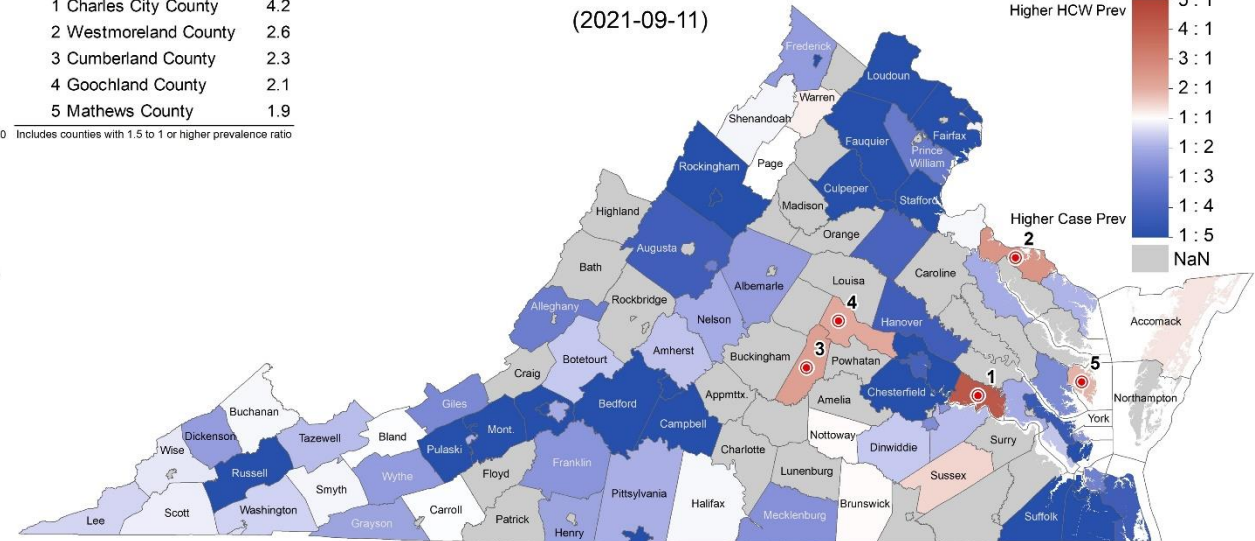
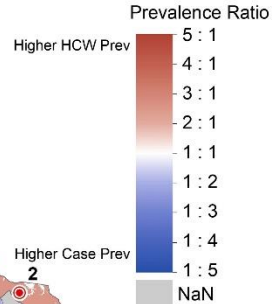
HCW Point Prevalence by Zip Code  
(2021-09-11)

Beyond  
Scale



## HCW to Public Prevalence Ratio

HCW Prevalence / Case Prevalence  
(2021-09-11)



Note: Scale differs from general public prevalence maps.



# Current Hot-Spots

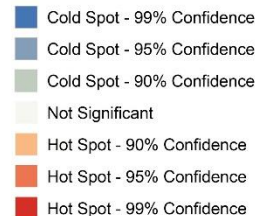
## Case rates that are significantly different from neighboring areas or model projections

- **Spatial:** Getis-Ord Gi\* based hot spots compare clusters of zip codes with weekly case prevalence higher than nearby zip codes to identify larger areas with statistically significant deviations
- **Temporal:** The weekly case rate (per 100K) projected last week compared to observed by county, which highlights temporal fluctuations that differ from the model's projections

### Spatial Hotspots

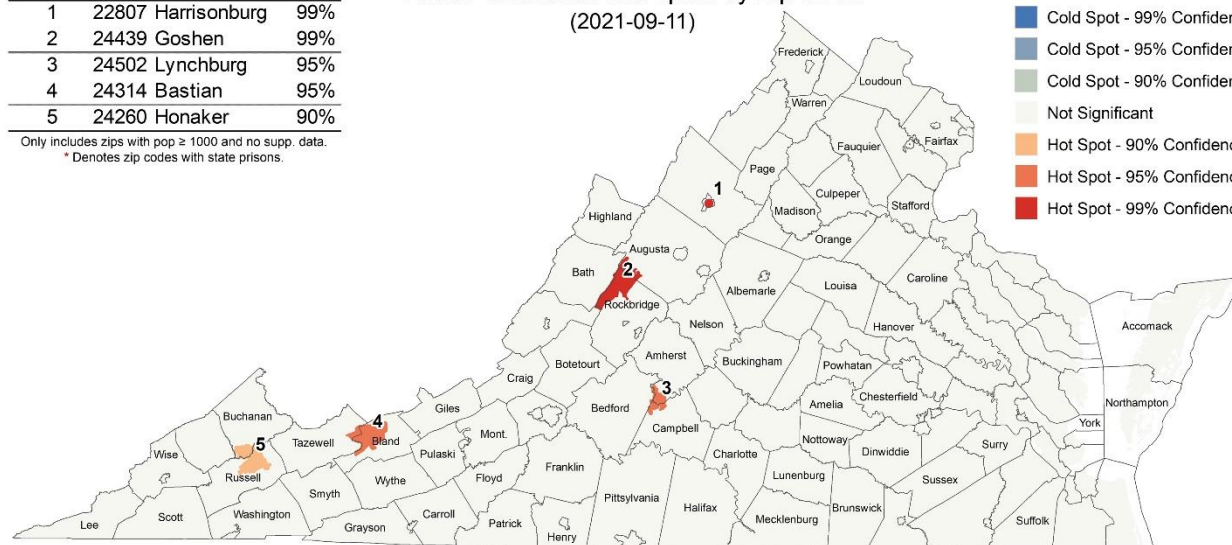
Point Prevalence Hot Spots by Zip Code  
(2021-09-11)

Getis-Ord Gi\* HotSpots



| Spot | Zip Code | Name         | Conf. |
|------|----------|--------------|-------|
| 1    | 22807    | Harrisonburg | 99%   |
| 2    | 24439    | Goshen       | 99%   |
| 3    | 24502    | Lynchburg    | 95%   |
| 4    | 24314    | Bastian      | 95%   |
| 5    | 24260    | Honaker      | 90%   |

Only includes zips with pop ≥ 1000 and no supp. data.  
\* Denotes zip codes with state prisons.

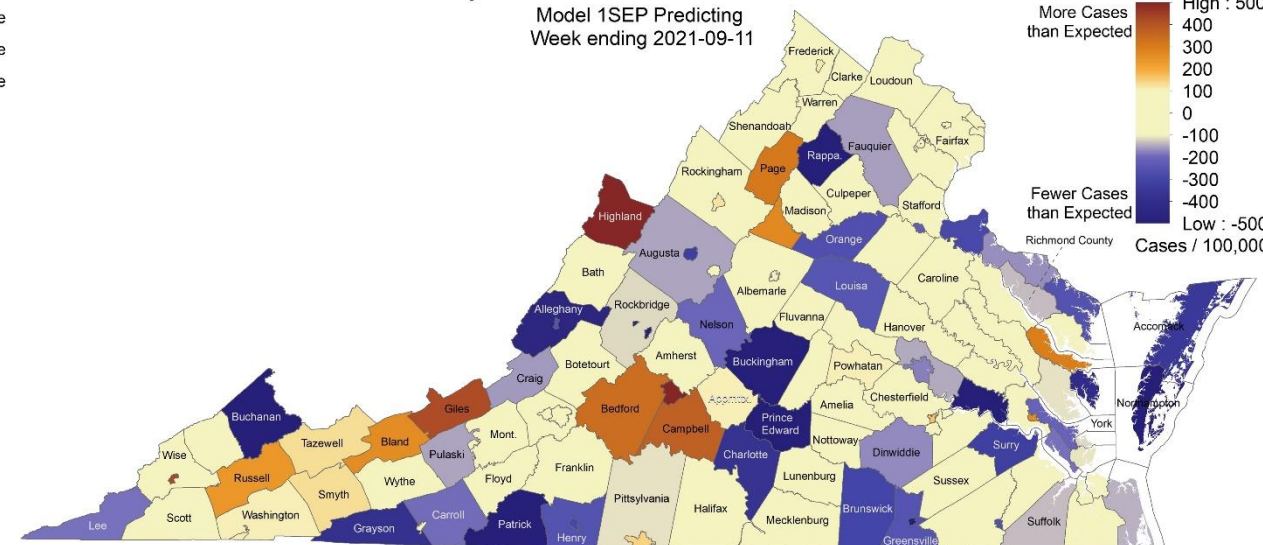
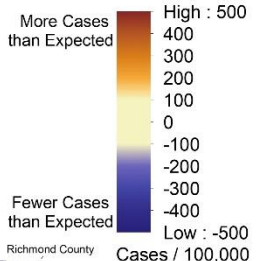


Based on Global Empirical Bayes smoothed point prevalence for week ending 2021-09-11.

### Clustered Temporal Hotspots

Weekly Point Prevalence Model Residuals  
Model 1SEP Predicting  
Week ending 2021-09-11

Residual



Moran's I = 0.030371, Z-Score = 1.678609, P-Value = 0.093228  
No Residual Autocorrelation Detected

# Model Update – Adaptive Fitting

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# Adaptive Fitting Approach

## Each county fit precisely, with recent trends used for future projection

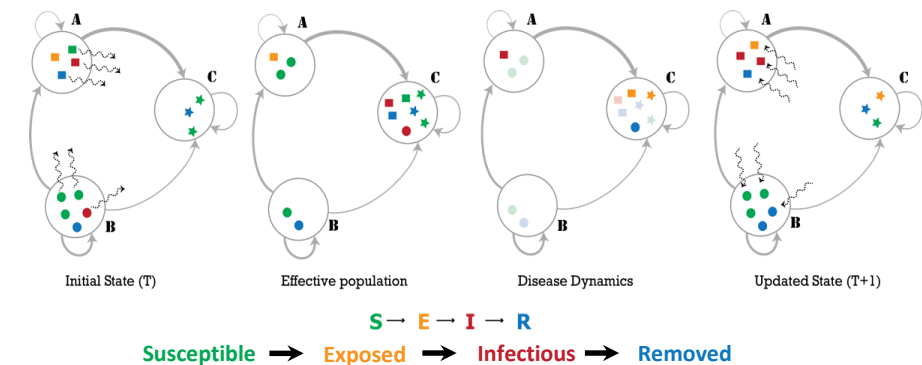
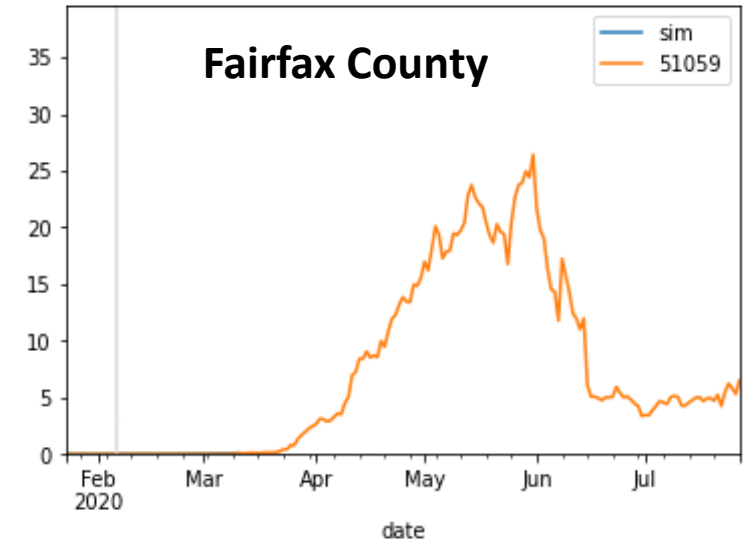
- Allows history to be precisely captured, and used to guide bounds on projections

## Model: An alternative use of the same meta-population model, PatchSim

- Allows for future “what-if” Scenarios to be layered on top of calibrated model
- Eliminates connectivity between patches, to allow calibration to capture the increasingly unsynchronized epidemic

## External Seeding: Steady low-level importation

- Widespread pandemic eliminates sensitivity to initial conditions
- Uses steady 1 case per 10M population per day external seeding



# Using Ensemble Model to Guide Projections

Ensemble methodology that combines the Adaptive with machine learning and statistical models such as:

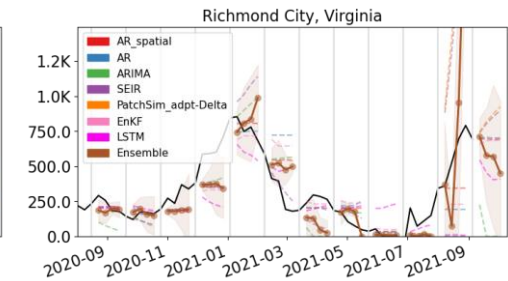
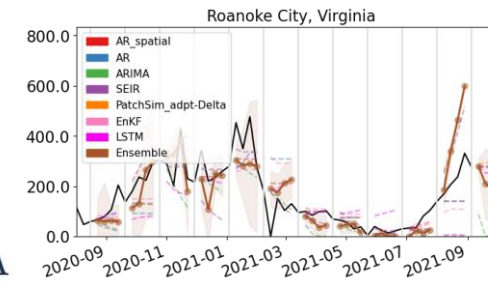
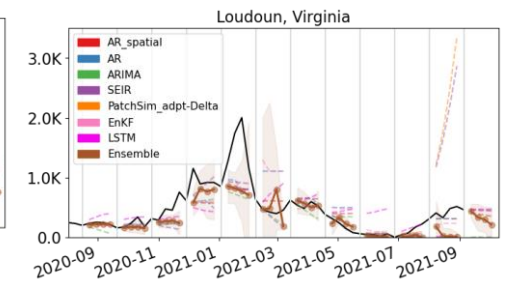
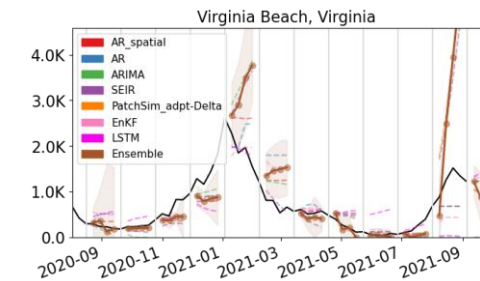
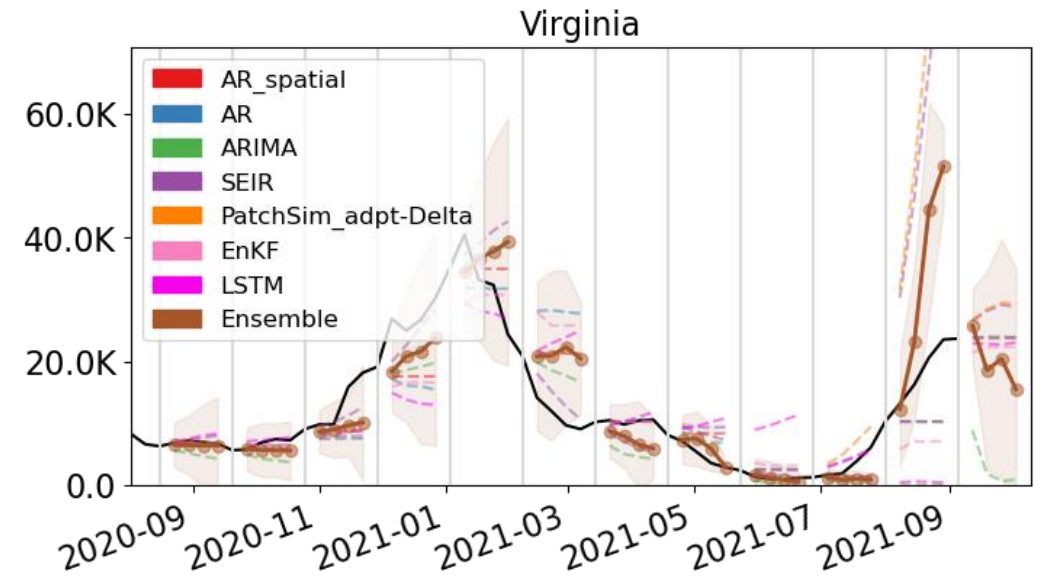
- Autoregressive (AR, ARIMA)
- Neural networks (LSTM)
- Kalman filtering (EnKF)

Weekly forecasts done at county level.

Models chosen because of their track record in disease forecasting and to increase diversity and robustness.

Ensemble forecast provides additional 'surveillance' for making scenario-based projections.

Also submitted to CDC Forecast Hub.



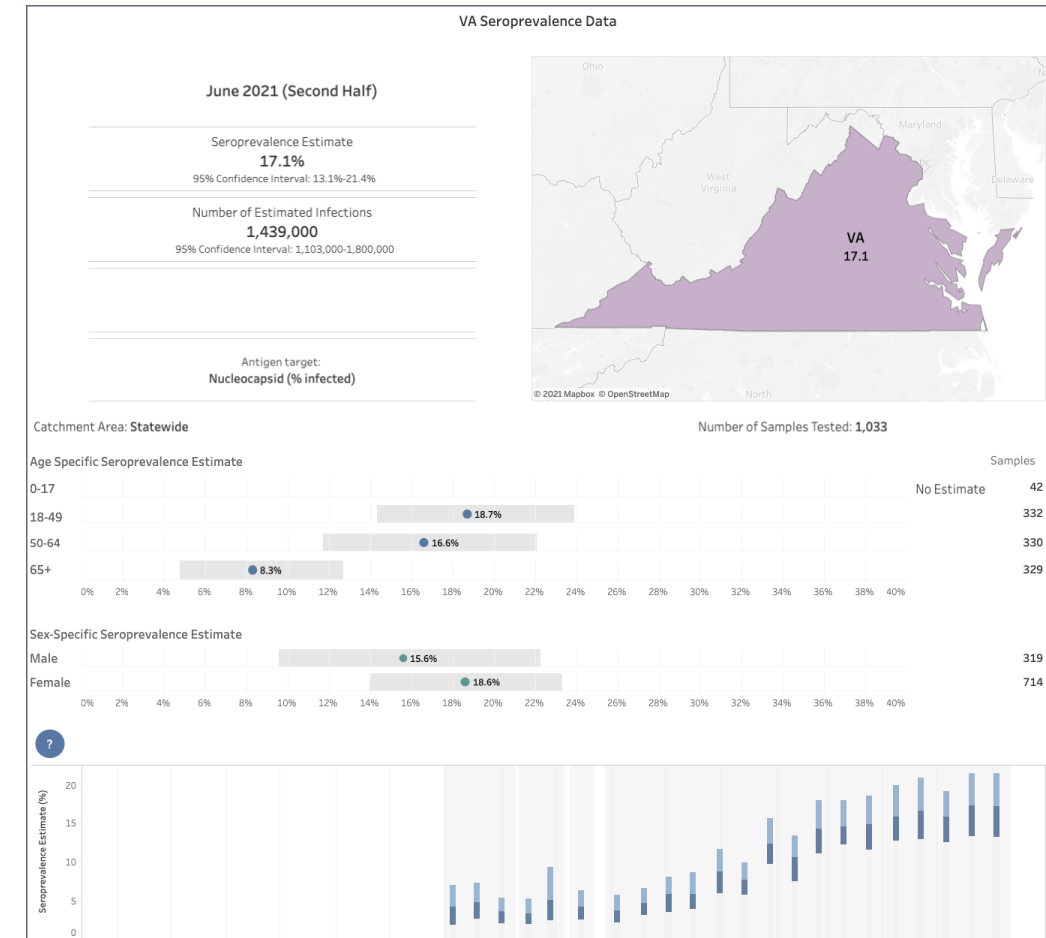
# Seroprevalence updates to model design

**Several seroprevalence studies provide better picture of how many actual infections have occurred**

- CDC Nationwide Commercial Laboratory Seroprevalence Survey

**These findings are equivalent to an ascertainment ratio of ~2x in the future, with bounds of (1.3x to 3x)**

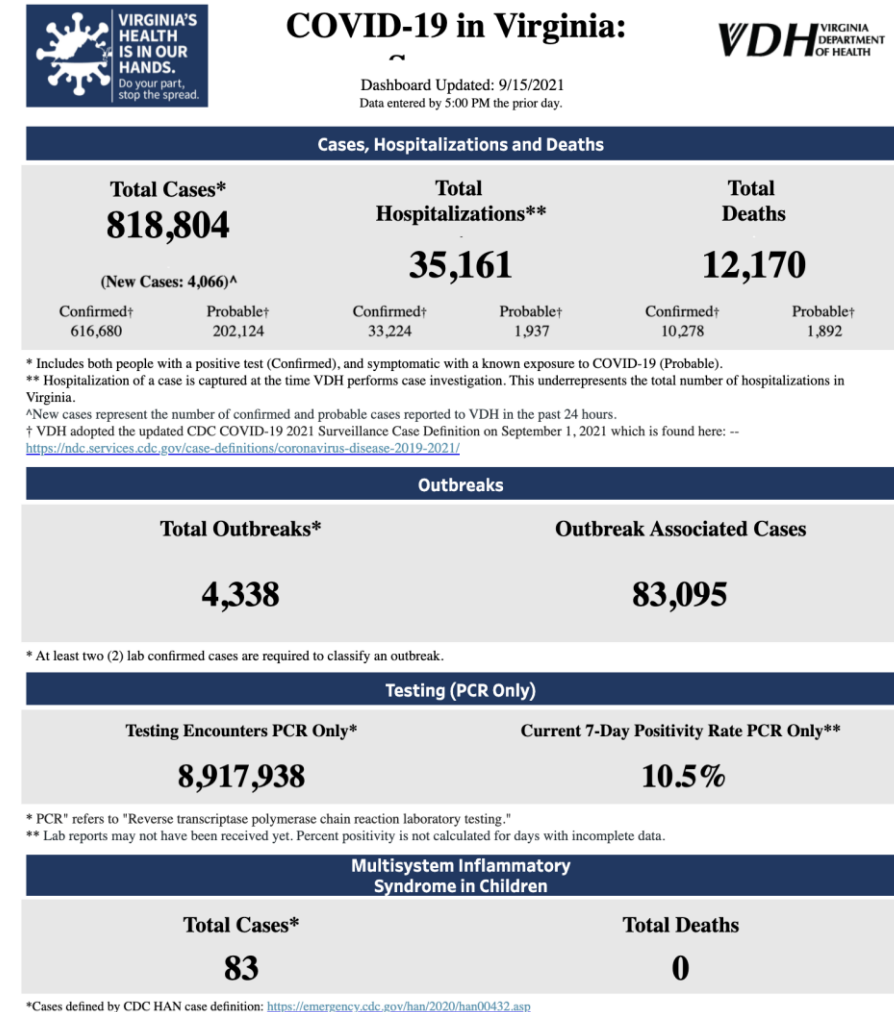
- Thus for 2x there are 2 total infections in the population for every confirmed case recently
- This measure now fully tracks the estimated ascertainment over time
- Uncertainty design has been shifted to these bounds (previously higher ascertainments as was consistent earlier in the pandemic were being used)



<https://covid.cdc.gov/covid-data-tracker/#national-lab>

# Calibration Approach

- **Data:**
  - County level case counts by date of onset (from VDH)
  - Confirmed cases for model fitting
- **Calibration:** fit model to observed data and ensemble's forecast
  - Tune transmissibility across ranges of:
    - Duration of incubation (5-9 days), infectiousness (3-7 days)
    - Undocumented case rate (1x to 7x) guided by seroprevalence studies
    - Detection delay: exposure to confirmation (4-12 days)
  - Approach captures uncertainty, but allows model to precisely track the full trajectory of the outbreak
- **Project:** future cases and outcomes generated using the collection of fit models run into the future
  - **Mean trend from last 14 days of observed cases and first week of ensemble's forecast used**
  - Outliers removed based on variances in the previous 3 weeks
  - 2 week interpolation to smooth transitions in rapidly changing trajectories
- **Outcomes:** Data driven by shift and ratio that has least error in last month of observations
  - Hospitalizations: 3 days from confirmation, 6.8% of cases hospitalized
  - Deaths: 11 days from confirmation, 1.45% of cases die



Accessed 9:45am September 15, 2021  
<https://www.vdh.virginia.gov/coronavirus/>

# Scenarios – Transmission Conditions

- Variety of factors continue to drive transmission rates
  - Seasonal impact of weather patterns, travel and gatherings, fatigue and premature relaxation of infection control practices
- **Waning Immunity:** Mean of one year protection (rate of 0.0027) similar to [Pfizer study](#)
- **Projection Scenarios:**
  - **Adaptive:** Control remains as is currently experienced into the future with assumption that Delta remains as the majority strain
  - **Adaptive-SeptSurge:** Following Labor Day 2021, transmission rates return to the median level from Sept-Nov of 2020 with a 60% boost over ancestral strain that dominated then.
  - **Adaptive-Surge Control:** Starting in one week behaviors and mitigation efforts ramp up over a 2-week period culminating in a 25% reduction in transmission
  - **Adaptive-Fall:** Control remains as is currently experienced into the future, with an increase in transmission that is 60% stronger than the median experienced October 2020 through February 2021 starting on Nov 1st

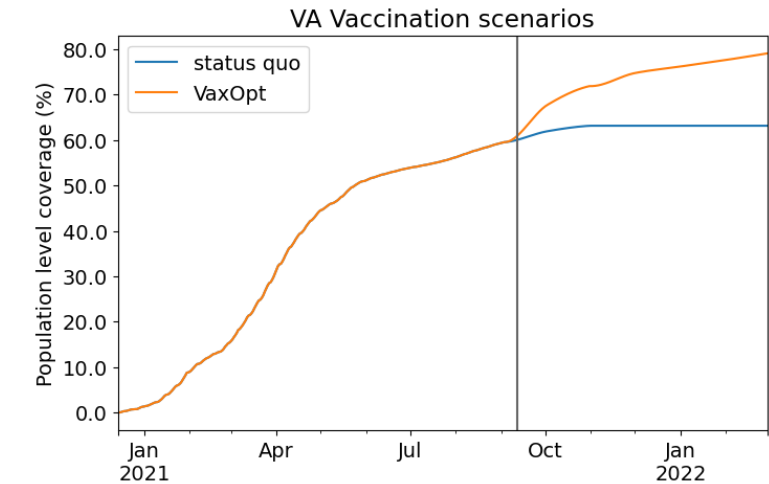
# Scenarios – Vaccination Conditions

## Vaccine Characteristics

- **Pfizer/Moderna:** 50% after first dose, 95% after second dose (3.5 week gap) **J & J :** 67% efficacy after first dose
- Delay to efficacy from doses is 14 days, immunity lasts at least 7m ([NEJM study](#))

## Vaccine Administration Scenarios

- **Status quo (no label):** COVIDcast corrected acceptance estimates (statewide mean is ~80% adults, 65% of population) reached by end of October.
- **Optimistic (VaxOpt):** Expand VA mean acceptance to include “probably not” (~85% adults) with addition of childhood (5-11 yo) rollout starting in Nov 1<sup>st</sup>. This follows the same rates as observed of adolescents and results in a net increase of ~10% of population by end of February. Additionally, all counties guaranteed to reach a minimum of 65%, max of 95% by end of October
- Acceptance at county level = regional acceptance +/- relative current vax
- Front-loaded rollout (two-thirds of the remaining in half the time)



| Monthly first doses | status quo |        |        | VaxOpt     |        |        |
|---------------------|------------|--------|--------|------------|--------|--------|
|                     | Date       |        |        | Date       |        |        |
|                     | 2020-12-31 | 109.1K | 109.1K | 2020-12-31 | 109.1K | 109.1K |
|                     | 2021-01-31 | 645.5K | 645.5K | 2021-01-31 | 754.6K | 754.6K |
|                     | 2021-02-28 | 558.4K | 558.4K | 2021-02-28 | 1.3M   | 1.3M   |
|                     | 2021-03-31 | 1.3M   | 1.3M   | 2021-03-31 | 2.6M   | 2.6M   |
|                     | 2021-04-30 | 1.2M   | 1.2M   | 2021-04-30 | 3.8M   | 3.8M   |
|                     | 2021-05-31 | 573.9K | 573.9K | 2021-05-31 | 4.4M   | 4.4M   |
|                     | 2021-06-30 | 242.3K | 242.3K | 2021-06-30 | 4.6M   | 4.6M   |
|                     | 2021-07-31 | 196.9K | 196.9K | 2021-07-31 | 4.8M   | 4.8M   |
| Cumulative          | Date       |        |        | Date       |        |        |
|                     | 2021-08-31 | 267.9K | 267.9K | 2021-08-31 | 5.1M   | 5.1M   |
|                     | 2021-09-30 | 215.0K | 676.5K | 2021-09-30 | 5.3M   | 5.7M   |
|                     | 2021-10-31 | 116.3K | 399.4K | 2021-10-31 | 5.4M   | 6.1M   |
|                     | 2021-11-30 | 0.0    | 240.2K | 2021-11-30 | 5.4M   | 6.4M   |
|                     | 2021-12-31 | 0.0    | 124.9K | 2021-12-31 | 5.4M   | 6.5M   |
|                     | 2022-01-31 | 0.0    | 122.8K | 2022-01-31 | 5.4M   | 6.6M   |
|                     | 2022-02-28 | 0.0    | 122.6K | 2022-02-28 | 5.4M   | 6.7M   |
|                     | 2022-03-31 | 0.0    | 4.7K   | 2022-03-31 | 5.4M   | 6.7M   |



# Projection Scenarios – Combined Conditions

| Name                  | Txm Controls | Vax | Description   |
|-----------------------|--------------|-----|---|
| Adaptive              | C            | SQ  | Likely trajectory based on conditions remaining similar to the current experience   |
| Adaptive-VaxOpt       | C            | VO  | Vaccination through October reaches an optimistically high level of expanded coverage (85%)   |
| Adaptive-Fall         | Fall         | SQ  | Same as Adaptive, with increased transmissibility driven by seasonality and/or another variant starting Nov 1 <sup>st</sup>                 |
| Adaptive-Fall-VaxOpt  | Fall         | VO  | Optimistically expand vaccination with increased transmissibility driven by seasonality and/or another variant starting Nov 1 <sup>st</sup> |
| Adaptive-SurgeControl | 25%          | SQ  | Transmission rates in the next month reduced through increased control from non-pharmaceutical interventions, with status quo vax and Delta |
| Adaptive-SeptSurge    | Sept         | SQ  | Transmission rates return to rates experienced in May 2021 with status quo vaccination and increasing prevalence of Delta                   |

## Transmission Controls:

C = Current levels persist into the future

25% = Transmission rates are reduced by 25% with a gradual introduction, concluding in 4 weeks

Fall = Current levels until Nov 1<sup>st</sup>, then the median level from Oct-Feb of last year with 60% variant boost

Sept = Transmission rates return to median of Sept-Nov 2020 with 60% boost following Labor Day

## Vaccinations:

SQ = Status quo acceptance leads to low rates of vaccination through the summer

VO = Vaccination acceptance optimistically expands with increased rates through the summer

# Model Results

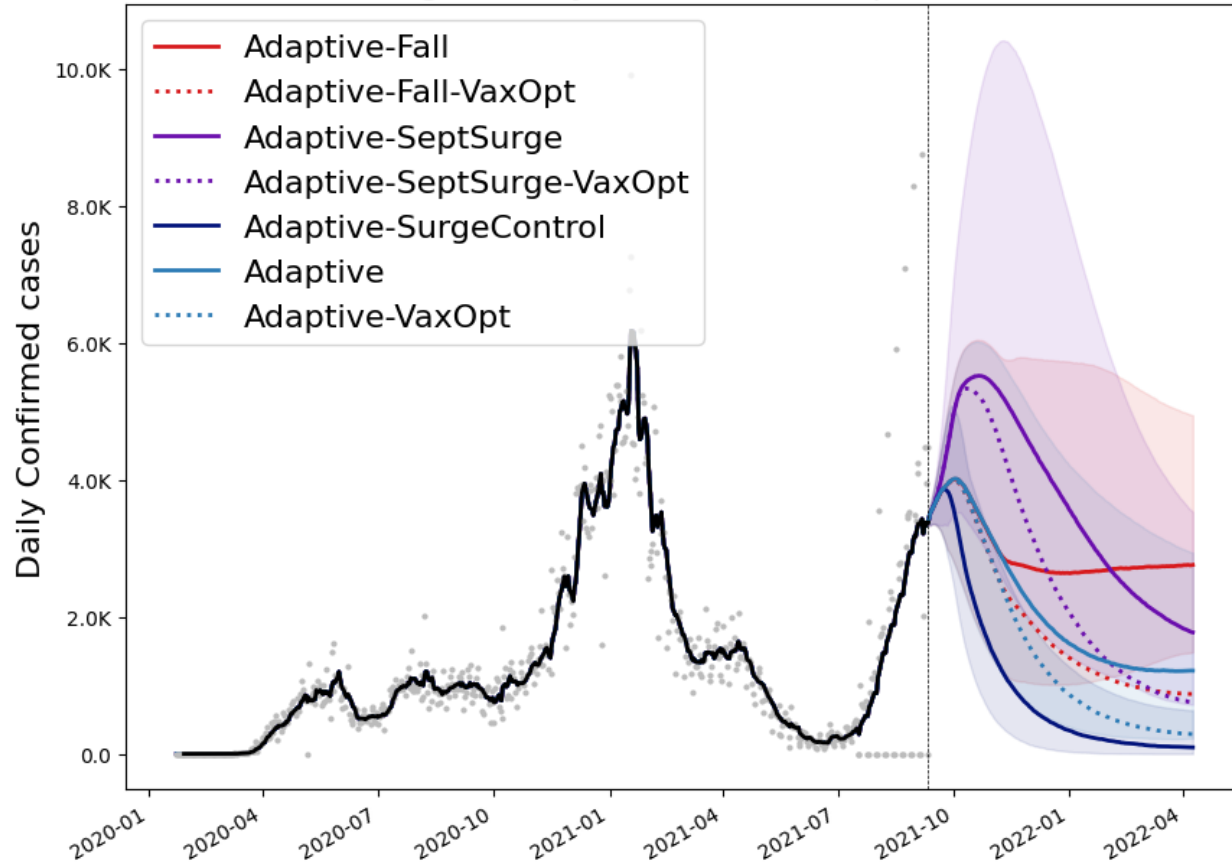
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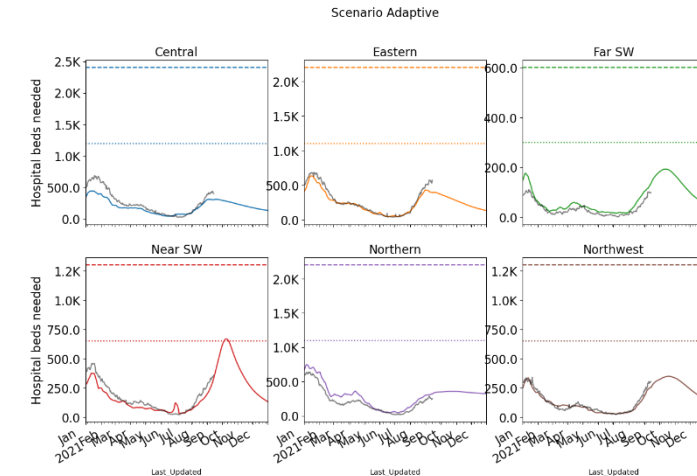
# Outcome Projections

## Confirmed cases

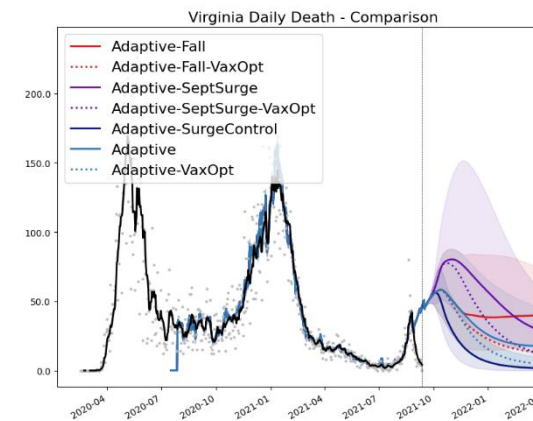
Virginia Daily Confirmed - Comparison



## Estimated Hospital Occupancy

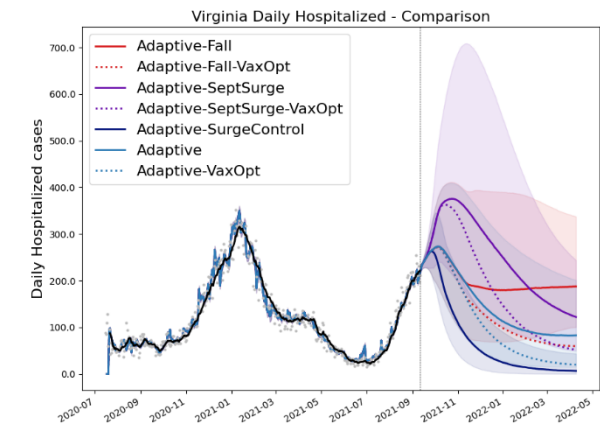


## Daily Deaths



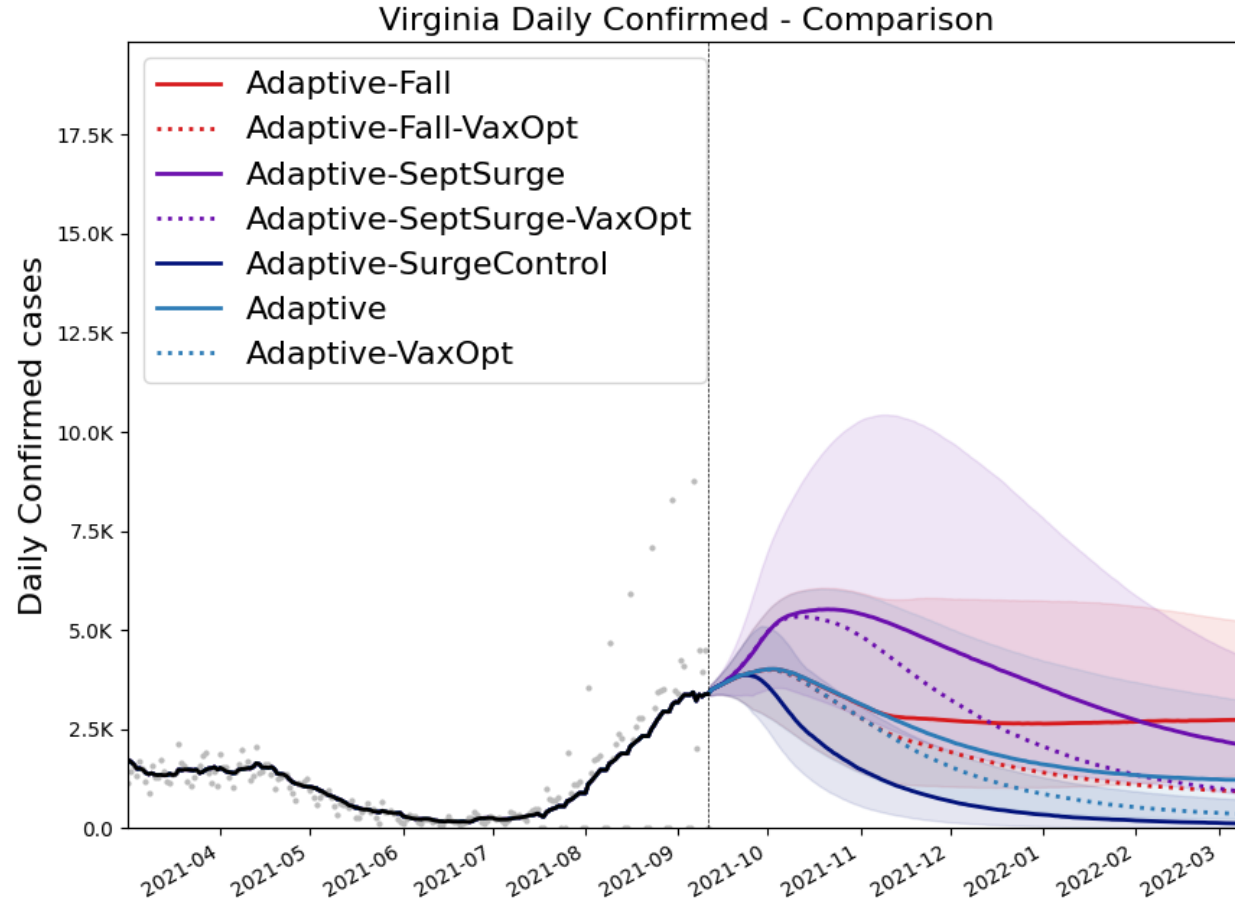
Death ground truth from VDH "Event Date" data, most recent dates are not complete

## Daily Hospitalized

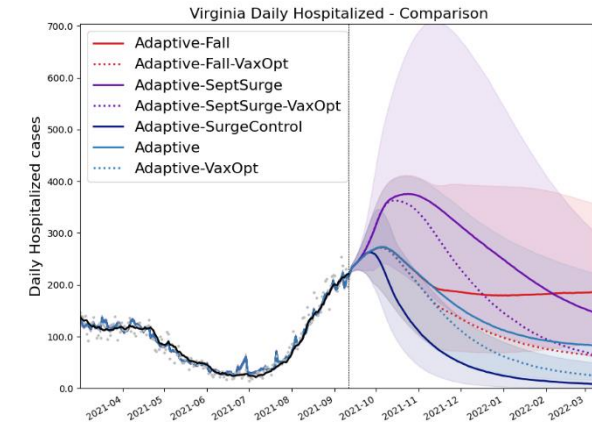


# Outcome Projections – Closer Look

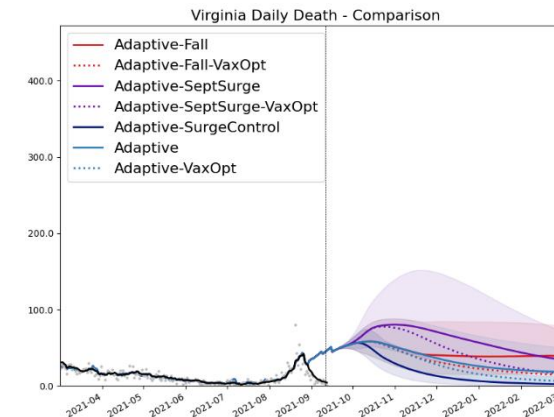
## Confirmed cases



## Daily Hospitalized



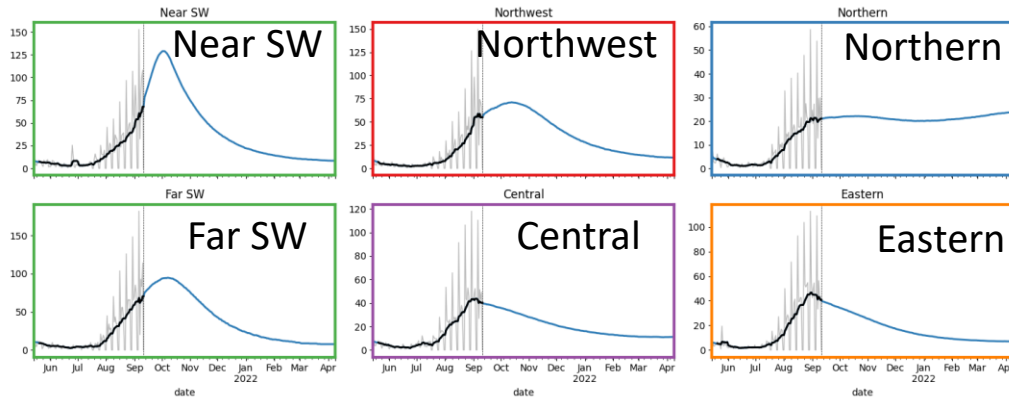
## Daily Deaths



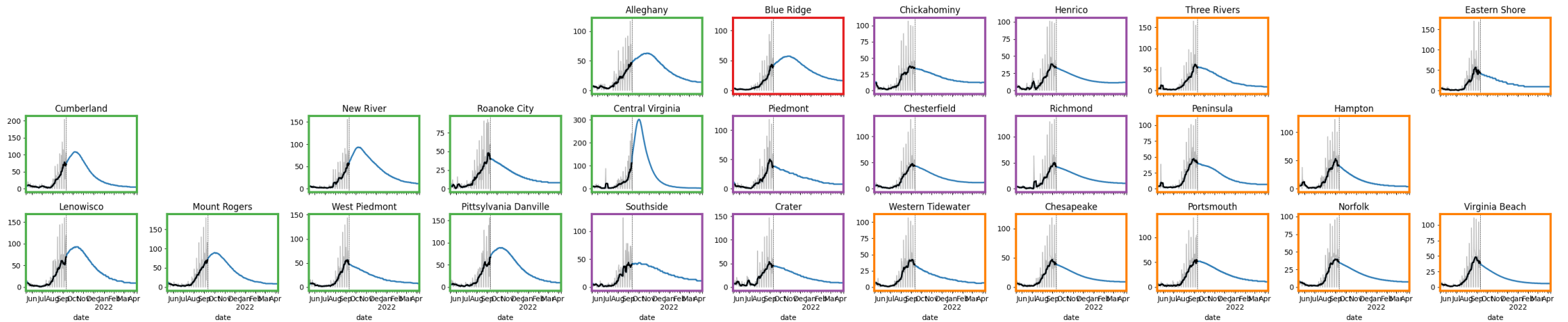
Death ground truth from VDH "Event Date" data, most recent dates are not complete

# District Level Projections: Adaptive

## Projections by Region



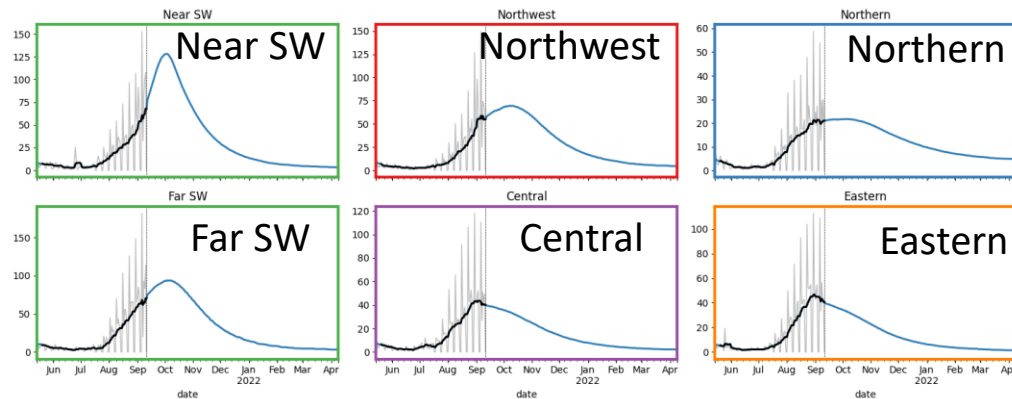
## Projections by District



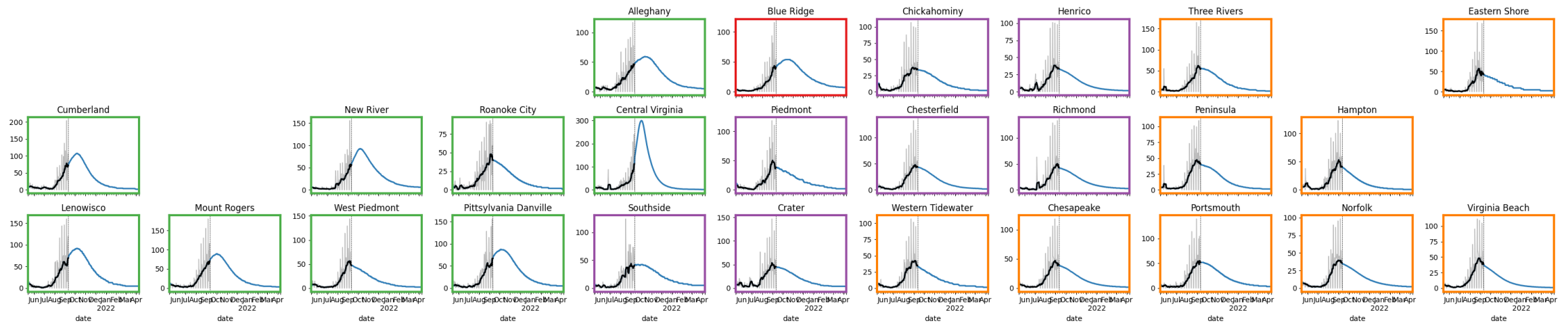
Daily confirmed cases rate (per 100K) by District (grey with 7-day average in black) with simulation colored by scenario

# District Level Projections: Adaptive-VaxOpt

## Projections by Region



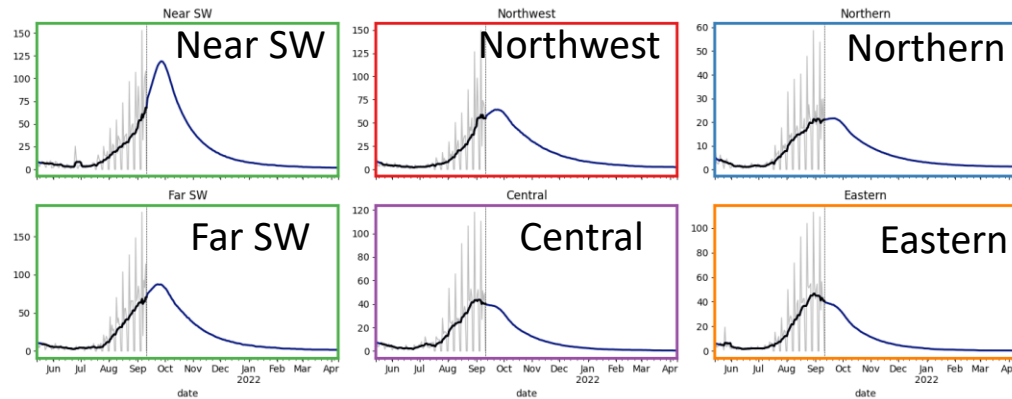
## Projections by District



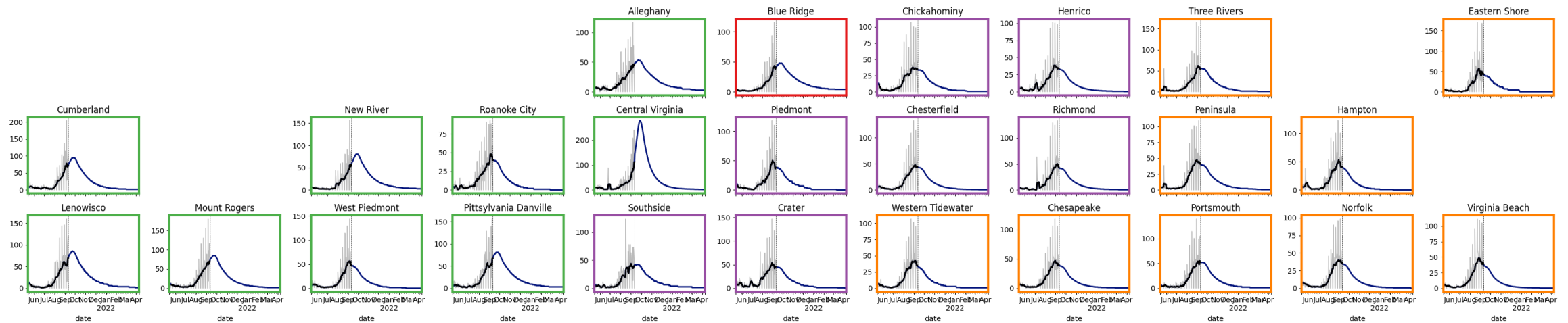
Daily confirmed cases rate (per 100K) by District (grey with 7-day average in black) with simulation colored by scenario

# District Level Projections: SurgeControl

## Projections by Region



## Projections by District

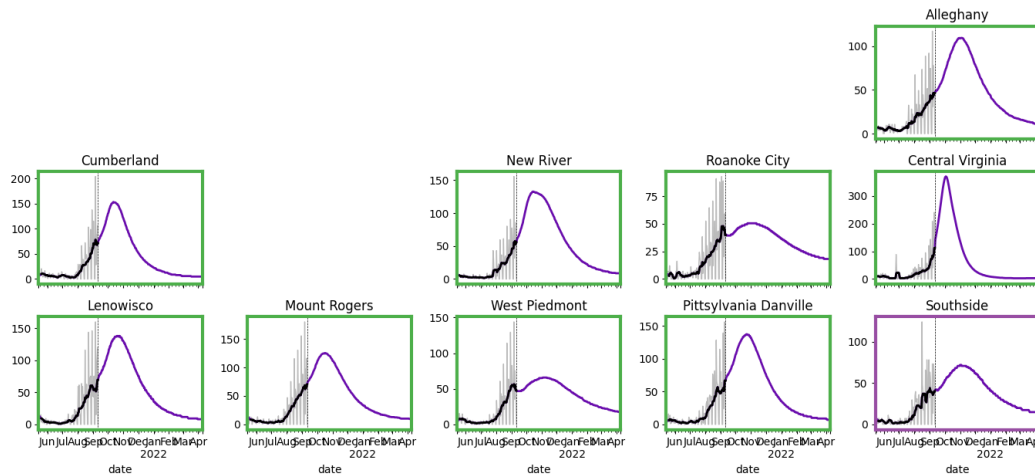
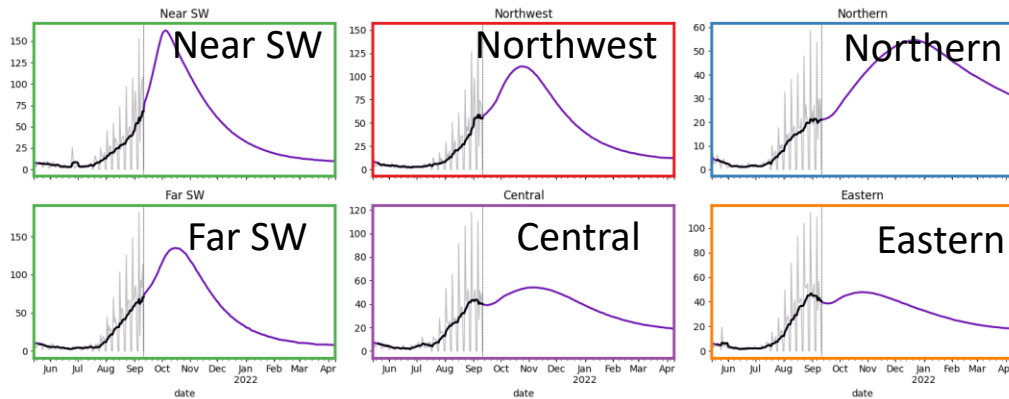


Daily confirmed cases rate (per 100K) by District (grey with 7-day average in black) with simulation colored by scenario

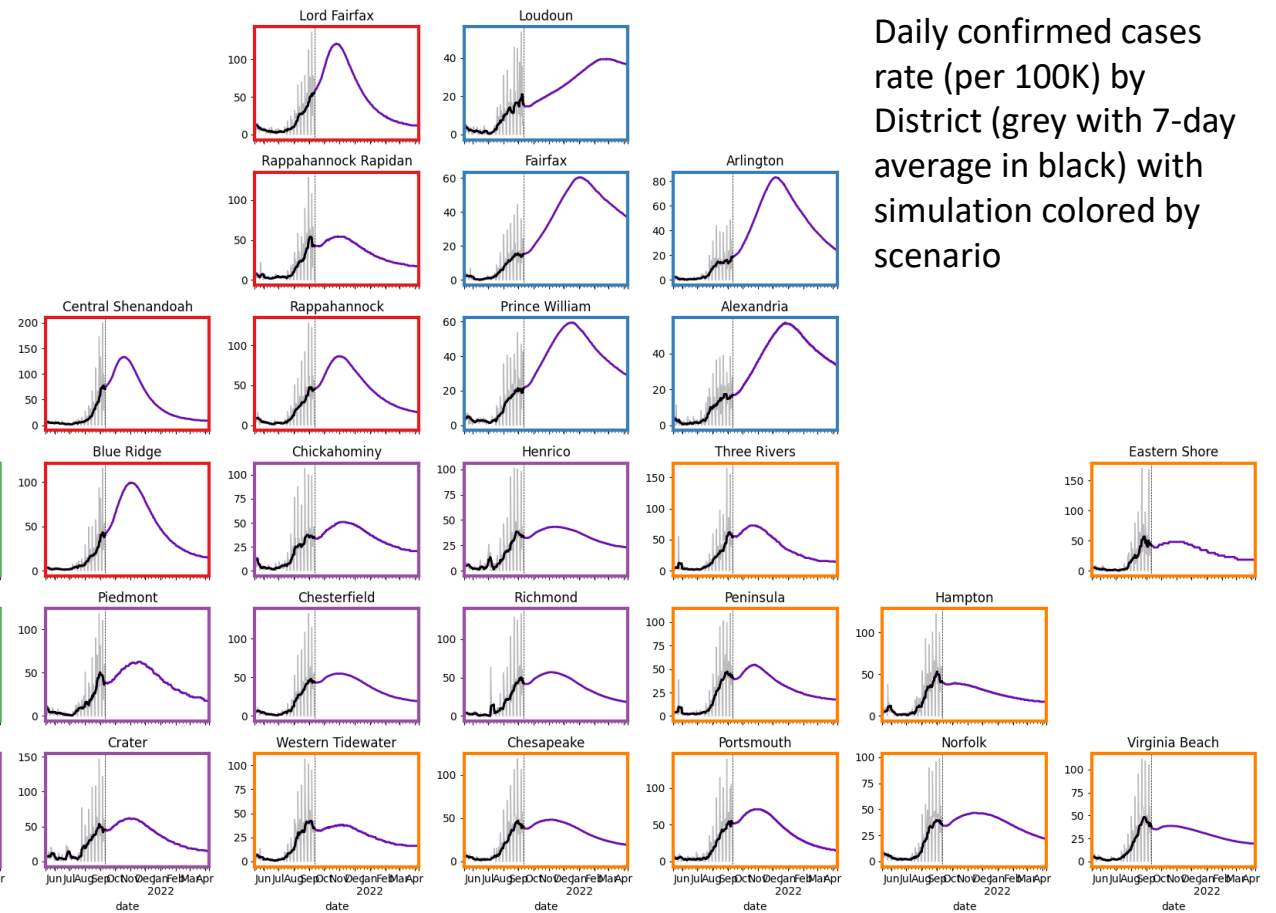


# District Level Projections: SeptSurge

## Projections by Region



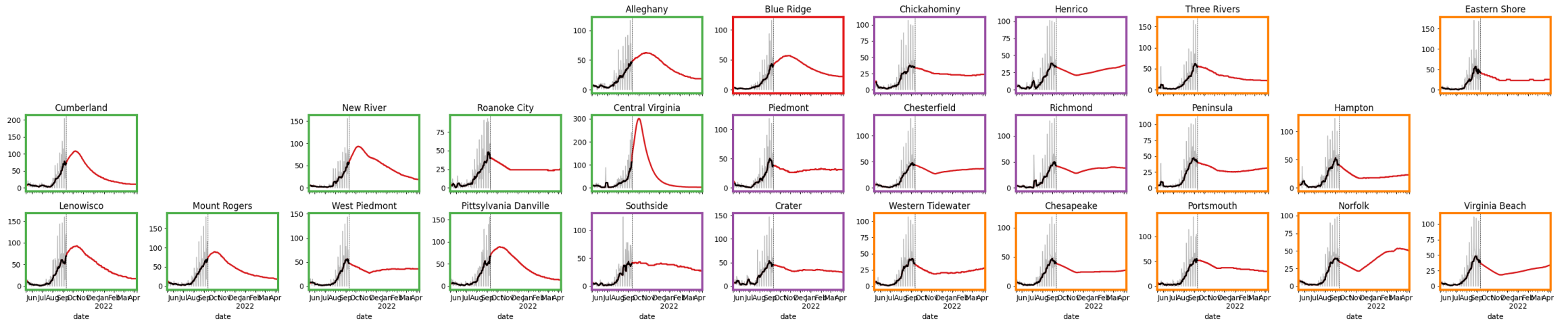
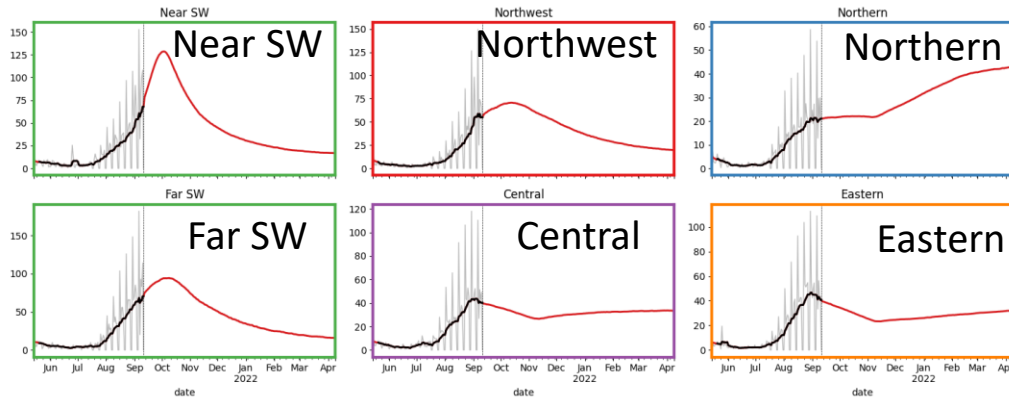
## Projections by District



Daily confirmed cases rate (per 100K) by District (grey with 7-day average in black) with simulation colored by scenario

# District Level Projections: Adaptive-Fall

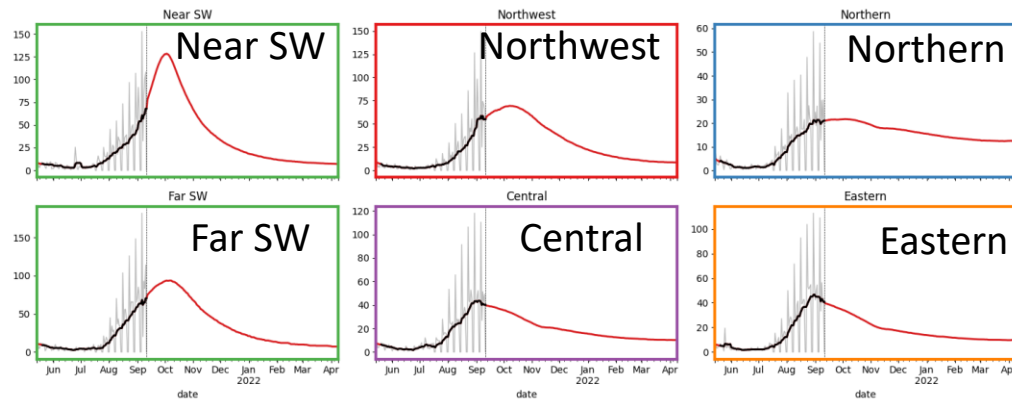
## Projections by Region



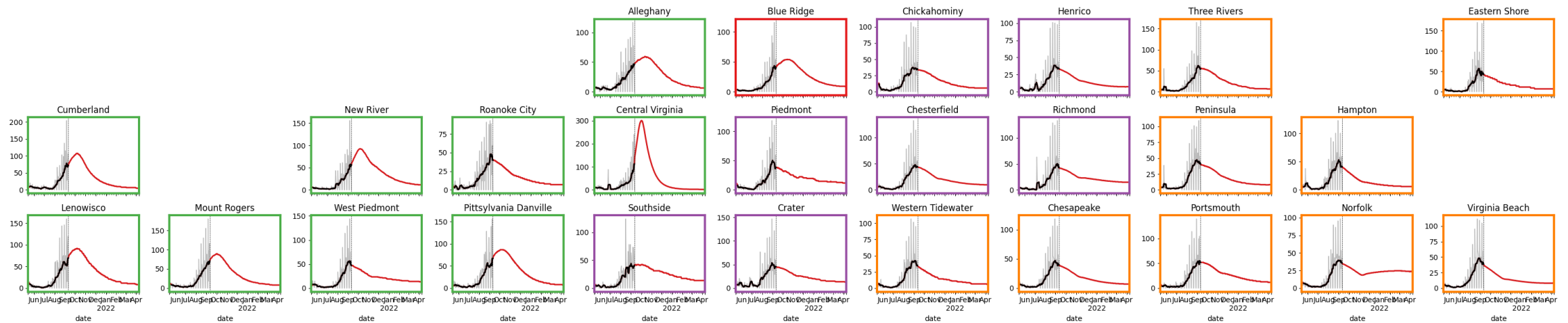
Daily confirmed cases rate (per 100K) by District (grey with 7-day average in black) with simulation colored by scenario

# District Level Projections: Adaptive-Fall-VaxOpt

## Projections by Region



## Projections by District



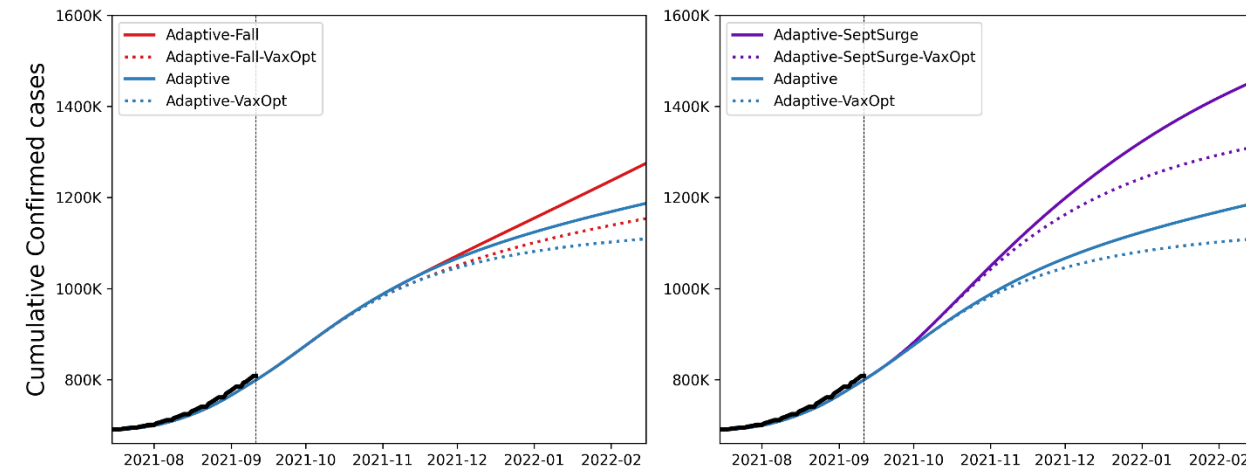
Daily confirmed cases rate (per 100K) by District (grey with 7-day average in black) with simulation colored by scenario

# Impact of expanded vaccine acceptance

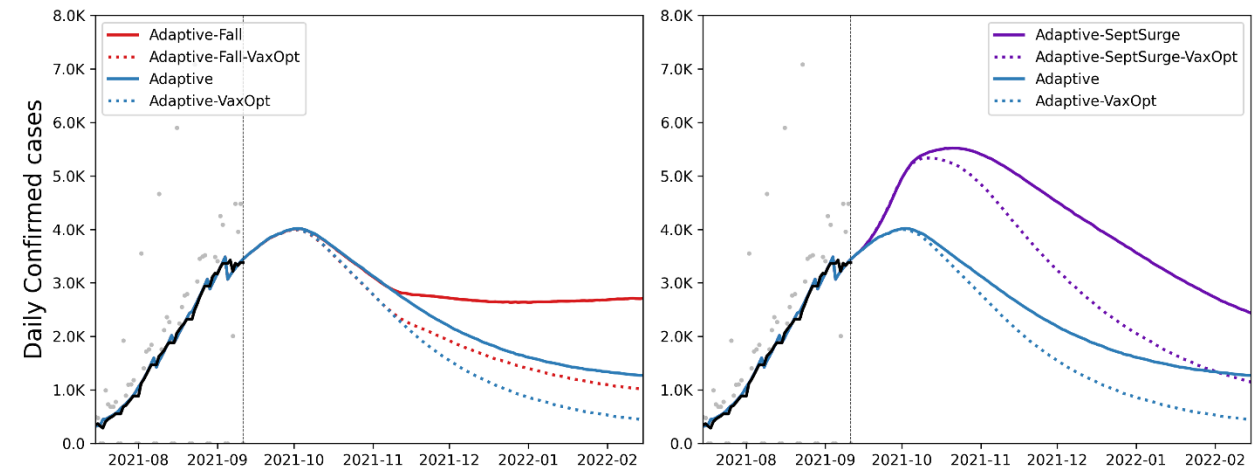
## Expanded Vax coverage with higher adult coverage & 5-11 year olds in Nov

- Even if transmission rates decline after a Delta wave, expanded vax coverage can reduce case counts by ~125K, in addition to providing further resilience to future waves
- A Fall Surge can slow the declining rates following the Delta wave
- Expanded vaccination coverage including children can further curtail the impact of a Fall Surge by up to ~215K cases or dampen the effects of a boosted SeptSurge by ~205K cases

Virginia Cumulative Confirmed - Comparison



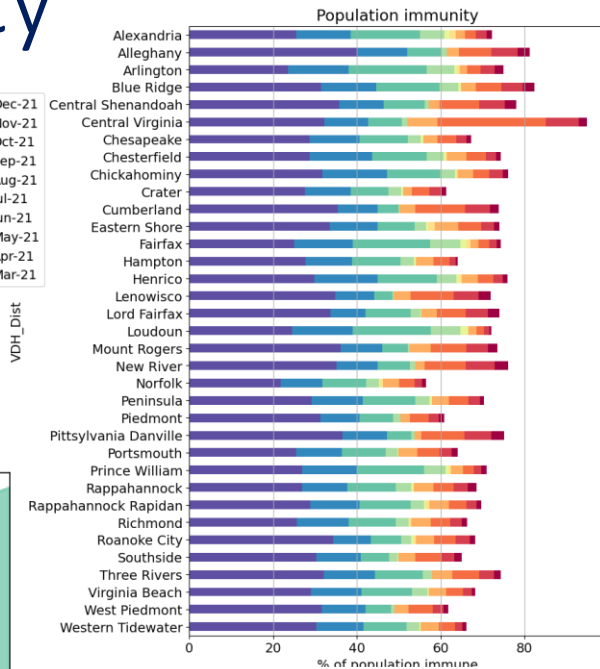
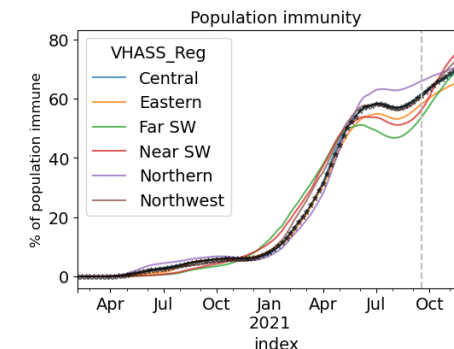
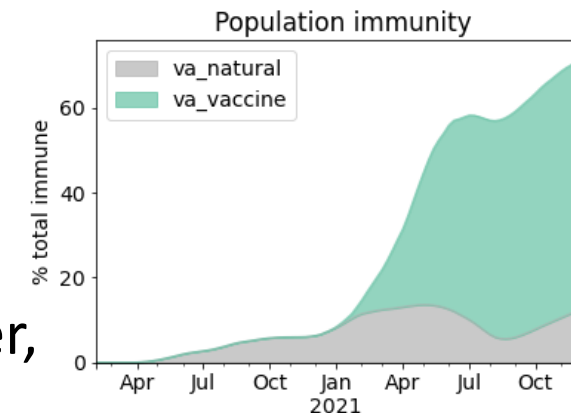
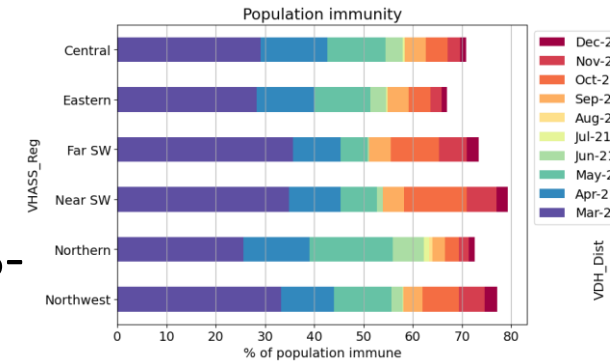
Virginia Daily Confirmed - Comparison



# Virginia's Progress on Population Immunity

## Natural Immunity and Vaccines combine to produce a population level of immunity

- Duration of immunity from infection with SARS-CoV2 still not well understood
  - We assume a conservative 6 month period of protection for these calculations
  - Do **not** factor in variant immune escape
  - Natural immunity is well calibrated to recent seroprevalence surveys
- Vaccine induced immunity is likely to last longer, we assume indefinite protection
  - This also assumes that all administered vaccines remain protective against current and future variants
- Population immunity depends on a very high proportion of the population getting vaccinated
  - Current models track measured seroprevalence



| Region    | % pop immune (est.)* |
|-----------|----------------------|
| Central   | 61%                  |
| Eastern   | 57%                  |
| Far SW    | 52%                  |
| Near SW   | 54%                  |
| Northern  | 65%                  |
| Northwest | 59%                  |
| Virginia  | 60%                  |

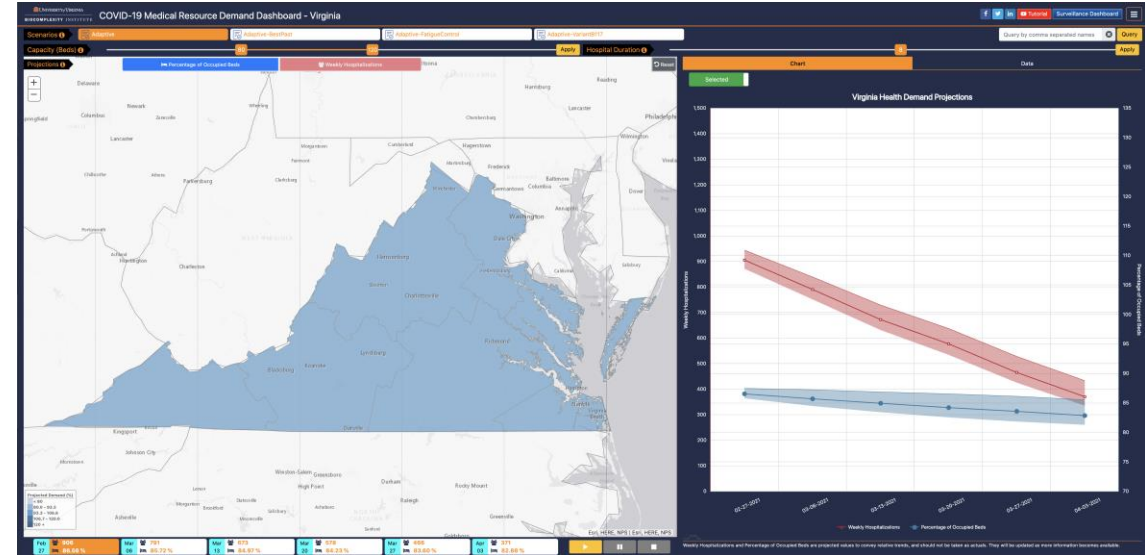
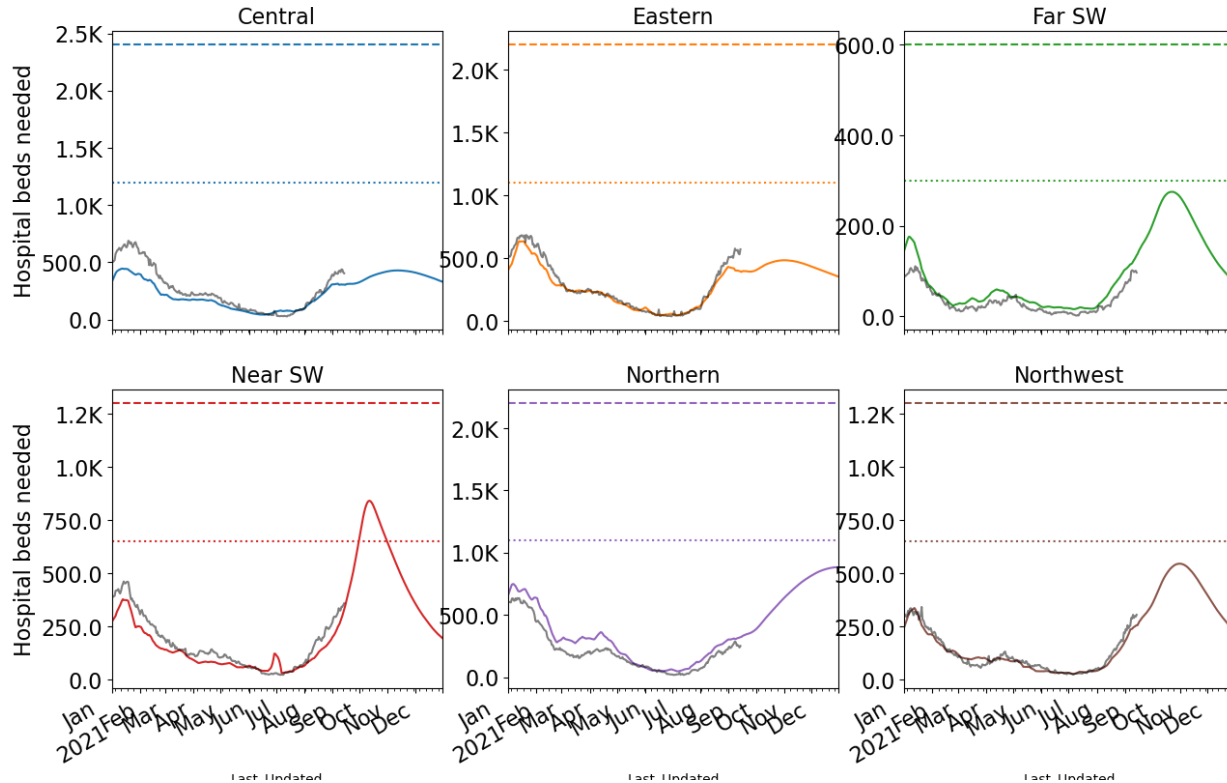
\* As of Sept 12, 2021 (updated to account for entire population)



# Hospital Demand and Bed Capacity by Region

## Capacities\* by Region – Adaptive SeptSurge

COVID-19 capacity ranges from 80% (dots) to 120% (dash) of total beds



<https://nssac.bii.virginia.edu/covid-19/vmrddash/>

**Adaptive SeptSurge scenario shows that even with Delta enhanced severity:**

- No regions should exceed their surge capacities
- Some regions may exceed initial capacities

\* Assumes average length of stay of 8 days

# Key Takeaways

Projecting future cases precisely is impossible and unnecessary.

Even without perfect projections, we can confidently draw conclusions:

- **Case rate growth in Virginia has slowed as many districts have started a decline from a peak, similarly across US many states have started declining from a peak; Case rates remain very high.**
- VA mean 7-day incidence is up at 43/100K from 38/100K, US also up at 50/100K (from 48/100K)
- Projections show reduced rate of increase and thus the impact has decreased compared to last week
- Recent updates:
  - Adjustment to higher levels of assumed immunity waning (natural and vaccine)
  - Added a SeptSurge based on transmission rates from last year Labor Day to Thanksgiving with variant boosting
  - Added Fall surge scenario to capture potential rebounds and further test immunity from expanded vaccination

The situation continues to change. Models continue to be updated regularly.

# Additional Analyses

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# Estimating Daily Reproductive Number – Redistributed weekend gap

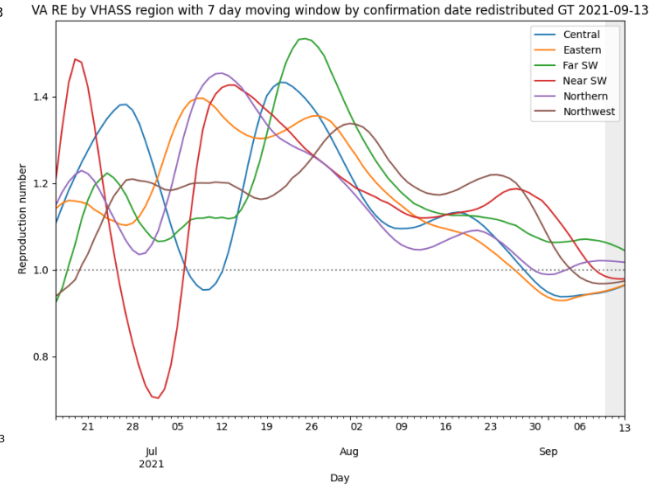
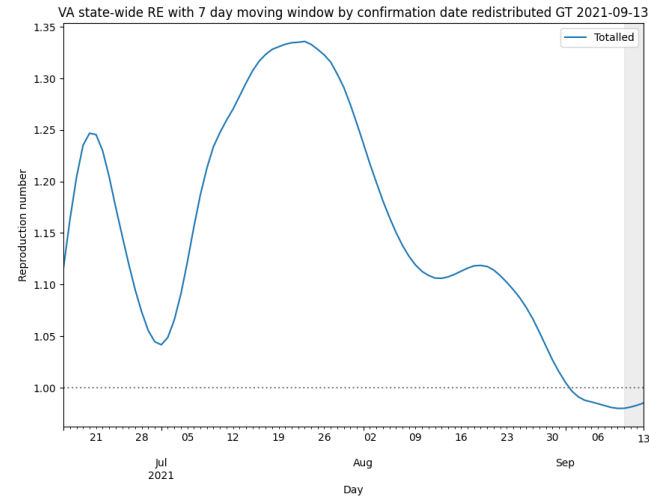
Sept 13<sup>th</sup> Estimates

| Region     | Date Confirmed<br>$R_e$ | Date Confirmed<br>Diff Last Week |
|------------|-------------------------|----------------------------------|
| State-wide | 0.984                   | -0.019                           |
| Central    | 0.942                   | 0.034                            |
| Eastern    | 0.939                   | 0.034                            |
| Far SW     | 1.070                   | 0.134                            |
| Near SW    | 1.043                   | 0.024                            |
| Northern   | 1.014                   | 0.100                            |
| Northwest  | 0.983                   | 0.001                            |

## Methodology

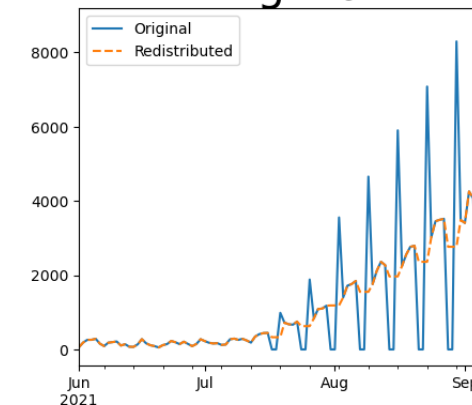
- Wallinga-Teunis method (EpiEstim<sup>1</sup>) for cases by confirmation date
- Serial interval: updated to discrete distribution from observations (mean=4.3, Flaxman et al, Nature 2020)
- Using Confirmation date since due to increasingly unstable estimates from onset date due to backfill

1. Anne Cori, Neil M. Ferguson, Christophe Fraser, Simon Cauchemez. A New Framework and Software to Estimate Time-Varying Reproduction Numbers During Epidemics. American Journal of Epidemiology, Volume 178, Issue 9, 1 November 2013, Pages 1505–1512, <https://doi.org/10.1093/aje/kwt133>



Skipping Weekend Reports biases estimates  
Redistributed Monday to fill in weekend, and then  
estimate R from "smoothed" time series

## Virginia



# Overview of relevant on-going studies

Other projects coordinated with CDC and VDH:

- **Scenario Modeling Hub:** Consortium of academic teams coordinated via MIDAS / CDC to that provides regular national projections based on timely scenarios
- **Genomic Surveillance:** Analyses of genomic sequencing data, VA surveillance data, and collaboration with VA DCLS to identify sample sizes needed to detect and track outbreaks driven by introduction of new variants etc.
- **Mobility Data driven Mobile Vaccine Clinic Site Selection:** Collaboration with VDH state and local, Stanford, and SafeGraph to leverage anonymized cell data to help identify



# COVID-19 Scenario Modeling Hub

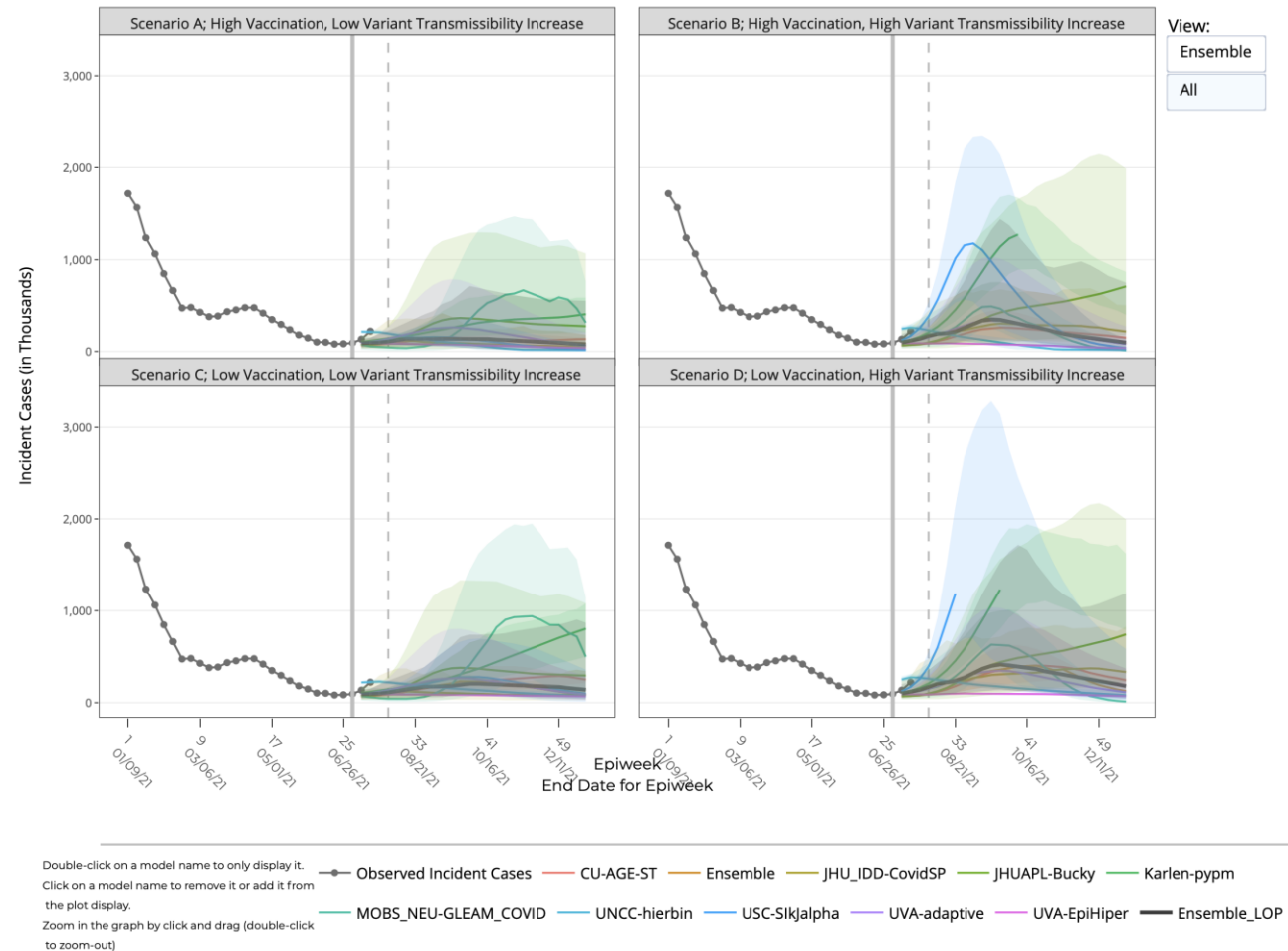
Collaboration of multiple academic teams to provide national and state-by-state level projections for 4 aligned scenarios that vary vaccine rates (high – low) and impact of the Delta variant (high and low)

- Round 8 in planning
- Round 7 now available

*Round 4 Results were published May 5<sup>th</sup>, 2021 in [MMWR](#)*

<https://covid19scenariomodelinghub.org/viz.html>

Projected Incident Cases by Epidemiological Week and by Scenario for Round 7  
( - Projection Epiweek; -- Current Week)



# COVID-19 Scenario Modeling Hub – Round 7

Round 7 scenarios explore the effects of a variant similar to Delta (B.1.617.2) against different backgrounds of vaccination. Includes some vax escape

## Vaccinations by Nov 30

- LowVacc – 70% overall coverage
- HighVacc – 80% overall coverage

## Emerging Variant Impact (5% prevalence on May 29<sup>th</sup>)

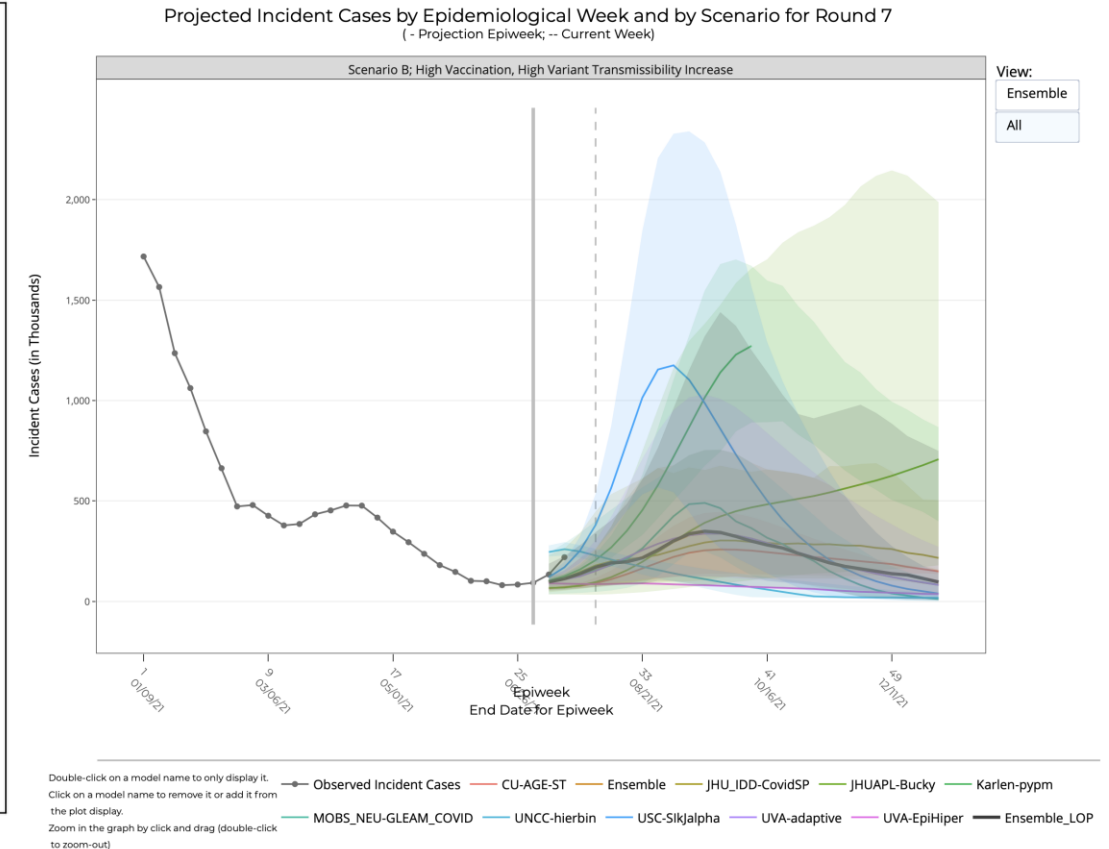
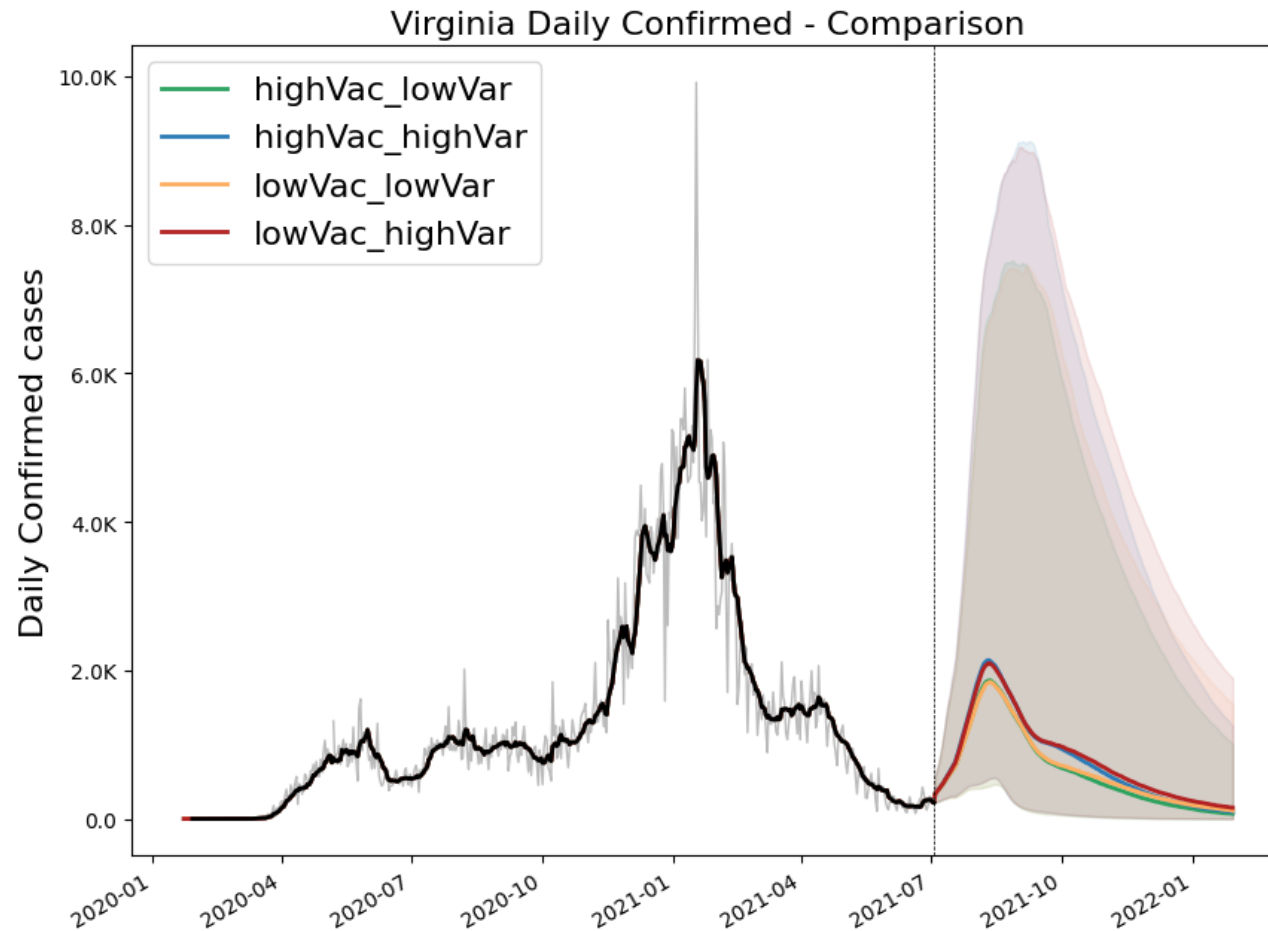
- LowVar – 40% more transmissible
- HighVar – 60% more transmissible

<https://covid19scenariomodelinghub.org/viz.html>

16-Sep-21

|   | LowVar  | HighVar   |
|---|---|---|
| See more detailed notes for each scenario below | Low Impact Variant<br>(low transmissibility increase, no immune escape)   | High Impact Variant<br>(high transmissibility increase, no immune escape)   |
| High Vaccination<br>(Low hesitancy)             | <b>Scenario A</b><br>Vaccination: <ul style="list-style-type: none"> <li>- Coverage saturates at <b>80% nationally</b> among the vaccine-eligible population* by December 31, 2021**</li> <li>- VE is <b>50%/90%</b> for Pfizer/Moderna against the Delta variant, against symptoms (1<sup>st</sup> /2<sup>nd</sup> dose)</li> <li>- J&amp;J no longer used</li> </ul> Variant: <ul style="list-style-type: none"> <li>- <b>40% increased transmissibility</b> as compared with Alpha for Delta variant. Initial prevalence estimated at state-level by teams.</li> </ul> | <b>Scenario B</b><br>Vaccination: <ul style="list-style-type: none"> <li>- Coverage saturates at <b>80% nationally</b> among the vaccine-eligible population* by December 31, 2021**</li> <li>- VE is <b>35%/85%</b> for Pfizer/Moderna against the Delta variant, against symptoms (1<sup>st</sup> /2<sup>nd</sup> dose)</li> <li>- J&amp;J no longer used</li> </ul> Variant: <ul style="list-style-type: none"> <li>- <b>60% increased transmissibility</b> as compared with Alpha for Delta variant. Initial prevalence estimated at state-level by teams.</li> </ul> |
| Low Vaccination<br>(High hesitancy)             | <b>Scenario C</b><br>Vaccination: <ul style="list-style-type: none"> <li>- Coverage saturates at <b>70% nationally</b> among the vaccine-eligible population* by December 31, 2021**</li> <li>- VE is <b>50%/90%</b> for Pfizer/Moderna against the Delta variant, against symptoms (1<sup>st</sup> /2<sup>nd</sup> dose)</li> <li>- J&amp;J no longer used</li> </ul> Variant: <ul style="list-style-type: none"> <li>- <b>40% increased transmissibility</b> as compared with Alpha for Delta variant. Initial prevalence estimated at state-level by teams.</li> </ul> | <b>Scenario D</b><br>Vaccination: <ul style="list-style-type: none"> <li>- Coverage saturates at <b>70% nationally</b> among the vaccine-eligible population* by December 31, 2021**</li> <li>- VE is <b>35%/85%</b> for Pfizer/Moderna against the Delta variant, against symptoms (1<sup>st</sup> /2<sup>nd</sup> dose)</li> <li>- J&amp;J no longer used</li> </ul> Variant: <ul style="list-style-type: none"> <li>- <b>60% increased transmissibility</b> as compared with Alpha for Delta variant. Initial prevalence estimated at state-level by teams.</li> </ul> |

# Modeling Hub – Round 7 Prelim Results



# COVID-19 Scenario Modeling Hub – Round 8 (ongoing)

Round 8 scenarios targeted at exploring the effect of waning immunity (natural and vaccine-induced) and varying levels of protection after waning

## Waning Rates

- Slow – exp. waning with mean=3yrs
- Fast – exp. waning with mean=1yr
- No waning (Sc A) as baseline

## Protection after Waning

- Age stratified protection from infection
- 80% or 90% protection from hosp/death

**High  
Protection**

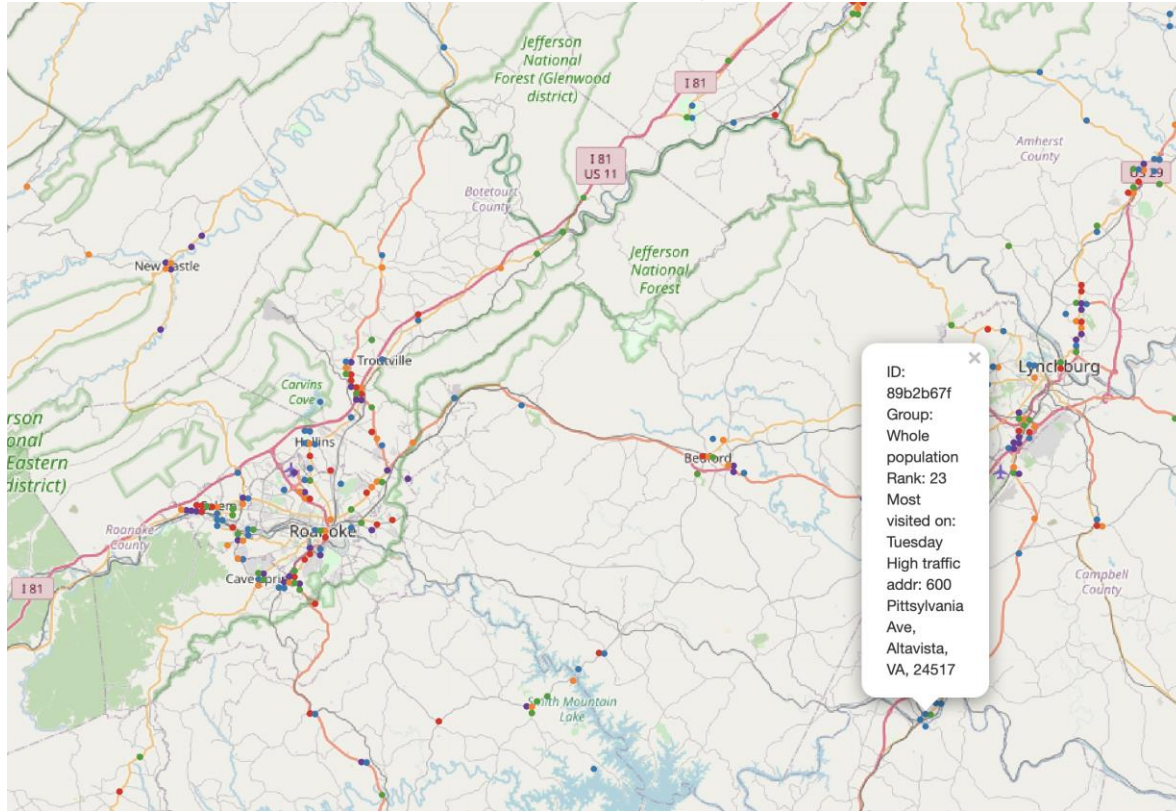
**Low  
Protection**

|  | Slow Waning   | Fast Waning  |
|--|---|--|
|  | See detailed notes on each scenario below   | <p><b>Slow waning of natural and vaccine-induced immunity</b><br/>(from no waning to exponential waning with mean of 3 yrs)</p> <p><b>Fast waning of natural and vaccine-induced immunity</b><br/>(exponential waning with mean of 1 year)</p>   |
|  | <p><b>Scenario A</b></p> <p>No Waning:</p> <ul style="list-style-type: none"> <li>- Vaccine-induced and natural immunity retain their initial protection throughout the simulation period</li> </ul>  | <p><b>Scenario B</b></p> <p>Waning:</p> <ul style="list-style-type: none"> <li>- <b>Exponentially distributed immune waning with mean of 1 year</b> (time to transition to partially immune state)</li> </ul> <p>In partially immune state:</p> <ul style="list-style-type: none"> <li>- Protection from infection is: <ul style="list-style-type: none"> <li>- 70% ≤ 65yrs</li> <li>- 35% &gt; 65yrs</li> </ul> </li> <li>- Protection from hospitalization and death is 90%</li> </ul> |
|  | <p><b>Scenario C</b></p> <p>Waning:</p> <ul style="list-style-type: none"> <li>- <b>Exponentially distributed immune waning with mean of 3 years</b> (time to transition to partially immune state)</li> </ul> <p>In partially immune state:</p> <ul style="list-style-type: none"> <li>- Protection from infection is: <ul style="list-style-type: none"> <li>- 50% ≤ 65yrs</li> <li>- 25% &gt; 65yrs</li> </ul> </li> <li>- Protection from hospitalization and death is 80%</li> </ul> | <p><b>Scenario D</b></p> <p>Waning:</p> <ul style="list-style-type: none"> <li>- <b>Exponentially distributed immune waning with mean of 1 year</b> (time to transition to partially immune state)</li> </ul> <p>In partially immune state:</p> <ul style="list-style-type: none"> <li>- Protection from infection is: <ul style="list-style-type: none"> <li>- 50% ≤ 65yrs</li> <li>- 25% &gt; 65yrs</li> </ul> </li> <li>- Protection from hospitalization and death is 80%</li> </ul> |
|  | <p><b>High protection against infection and severe disease after waning</b></p>   |  |
|  | <p><b>Low protection against infection and severe disease after waning</b></p>  |  |

<https://covid19scenariomodelinghub.org/>

# Data Recommended Mobile Vax Clinic Sites

## Detailed and Timely Locations



## Data Delivered and Disseminated to Locals

Provides a list of areas most visited by a given demographic group based on SafeGraph mobility data that links visits to specific sites and the home Census Block Group of the anonymized visitors

**Demographic Groups:** Black, Lantinx, Young Adults (20-40), Unvaccinated, and Whole Population

**Data Included:** Rank, Weight, most visited Day of Week, Highly Visited Address, and Lat-Long of area

**Goal:** Provide frequently visited locations based on populations and vaccination levels one desires to reach

**Example:** List of location in the Southside frequented by 20-40 year olds





# References

Venkatramanan, S., et al. "Optimizing spatial allocation of seasonal influenza vaccine under temporal constraints." *PLoS Computational Biology* 15.9 (2019): e1007111.

Arindam Fadikar, Dave Higdon, Jiangzhuo Chen, Bryan Lewis, Srinivasan Venkatramanan, and Madhav Marathe. Calibrating a stochastic, agent-based model using quantile-based emulation. *SIAM/ASA Journal on Uncertainty Quantification*, 6(4):1685–1706, 2018.

Adiga, Aniruddha, Srinivasan Venkatramanan, Akhil Peddireddy, et al. "Evaluating the impact of international airline suspensions on COVID-19 direct importation risk." *medRxiv* (2020)

NSSAC. PatchSim: Code for simulating the metapopulation SEIR model. <https://github.com/NSSAC/PatchSim>

Virginia Department of Health. COVID-19 in Virginia. <http://www.vdh.virginia.gov/coronavirus/>

Biocomplexity Institute. COVID-19 Surveillance Dashboard. <https://nssac.bii.virginia.edu/covid-19/dashboard/>

Google. COVID-19 community mobility reports. <https://www.google.com/covid19/mobility/>

Biocomplexity page for data and other resources related to COVID-19: <https://covid19.biocomplexity.virginia.edu/>

# Questions?

## Points of Contact

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## Biocomplexity COVID-19 Response Team

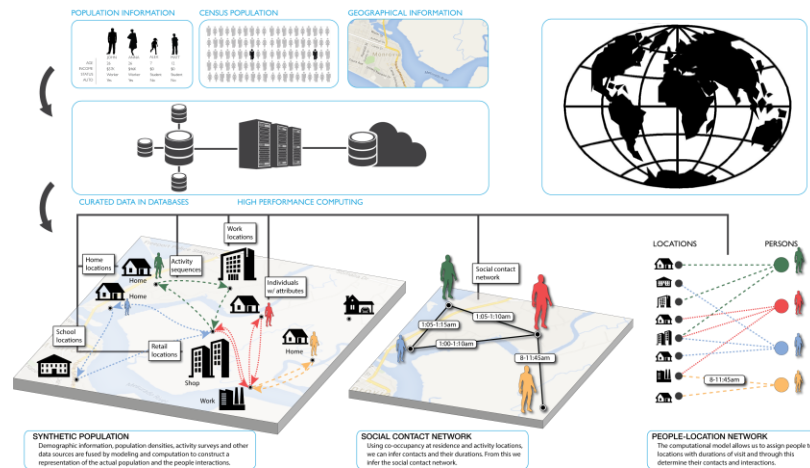
Aniruddha Adiga, Abhijin Adiga, Hannah Baek, Chris Barrett, Golda Barrow, Richard Beckman, Parantapa Bhattacharya, Jiangzhuo Chen, Clark Cucinell, Patrick Corbett, Allan Dickerman, Stephen Eubank, Stefan Hoops, Ben Hurt, Ron Kenyon, Brian Klahn, Bryan Lewis, Dustin Machi, Chunhong Mao, Achla Marathe, Madhav Marathe, Henning Mortveit, Mark Orr, Joseph Outten, Akhil Peddireddy, Przemyslaw Porebski, Erin Raymond, Jose Bayoan Santiago Calderon, James Schlitt, Samarth Swarup, Alex Telionis, Srinivasan Venkatramanan, Anil Vullikanti, James Walke, Andrew Warren, Amanda Wilson, Dawen Xie

# Supplemental Slides

# Agent-based Model (ABM )

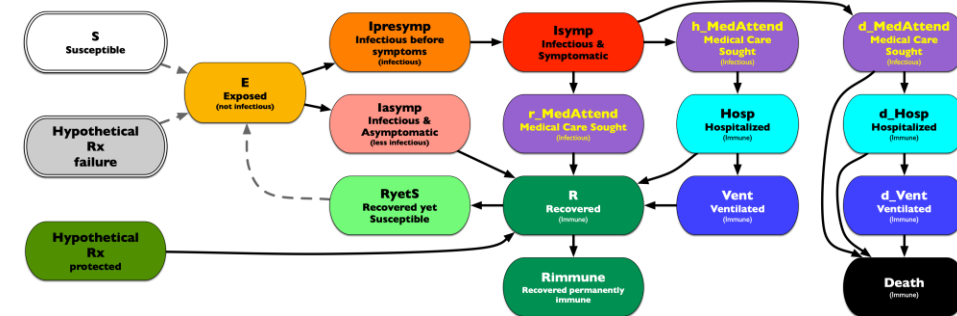
## EpiHiper: Distributed network-based stochastic disease transmission simulations

- Assess the impact on transmission under different conditions
- Assess the impacts of contact tracing



### Synthetic Population

- Census derived age and household structure
- Time-Use survey driven activities at appropriate locations



### Detailed Disease Course of COVID-19

- Literature based probabilities of outcomes with appropriate delays
- Varying levels of infectiousness
- Hypothetical treatments for future developments